

## Article

# An Output Feedback Controller for a Second-Order System Subject to Asymmetric Output Constraint Based on Lyapunov Function with Unlimited Domain

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**Abstract:** In this work, a new robust controller is designed for a second-order plant model, considering asymmetric output constraints. The tracking error convergence and output constraint are achieved by using a control law whose output feedback term is user-defined and bounded: it takes on large but finite and user-defined values for tracking error values equal to or higher than the constraint boundary, and it comprises a previously known user-defined function for tracking error values far from the constraint boundary. This is a significant contribution that remedies two important limitations of common output constraint control designs: the infinite control effort for tracking error equal to or higher than the constraint boundary, and the impossibility of using previously known user-defined functions in the output feedback function for tracking error values far from the constraint boundary. As another contribution, the control design is based on the dead-zone Lyapunov function, which facilitates the achievement of convergence to a compact set with user-defined size, avoidance of discontinuous signals in the controller, and robustness to model uncertainty or disturbances. The proposed output feedback term consists of the product between two functions of the tracking error, an increasing function and a sigmoid function, whose exact expressions are user-defined. Finally, the effectiveness of the developed controller is illustrated by the simulation of substrate concentration tracking in a continuous flow stirred bioreactor.

**Keywords:** second order system; asymmetric output constraint; unlimited domain Lyapunov function; dead-zone Lyapunov function; robust control

**MSC:** 37M99; 37M05



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## 1. Introduction

Many dynamic systems are subject to output constraint, which is related to performance requirements, physical limitations, or safety issues [1–3]. For instance, the joint position of robotic manipulators, and the velocity of nonholonomic and unmanned aerial vehicles should be limited [1,2,4].

In the control literature, a common strategy to tackle output constraint is the barrier Lyapunov function (BLF) [1], which was pioneered by [5]. Compared to other output constraint strategies, it has the advantage of easier implementation, as numerical or computationally intensive algorithms are avoided [4]. In the BLF strategy, the prescribed constant boundary defined for the tracking error is tackled through a Lyapunov function that involves a vertical asymptote at the prescribed boundary, so that it takes on large values for

the tracking error values close to the boundary and it is not defined for tracking error values exceeding it. The output feedback term of the resulting control law also exhibits these features. Therefore, the achievement of output constraint within the prescribed boundaries is a consequence of the enhanced control effort [1,6,7]. Usually, the Lyapunov function is logarithm or tangent type [4], whereas sliding mode control (SMC) [4,6] or backstepping control [2,8] are used as basic control design framework. The main improvements to the basic BLF control design for higher-order nonlinear systems are the consideration of asymmetric constraint [1–4]; consideration of no strict feedback form systems [1–3]; use of finite-time stabilization [3,9].

An alternative way to achieve output constraint is by using a Lyapunov function defined over the whole domain of the tracking error, and a controller with output feedback terms featuring large but bounded values for the tracking error equal to the prescribed boundary [10]. Some control designs with these features (which we call FOC features) are discussed in what follows. In [10], a modification is proposed for the Lyapunov function used in traditional prescribed performance control (PPC). In [11], an output-constrained SMC is designed for systems of relative degree two involving disturbances. The first derivative of the output is constrained but not the output. The controller guarantees that the constraints can be violated during a finite time, which can be shortened by properly choosing the controller parameters. When there is a constraint violation event, the control law includes an additive term that is the function of the error between the constrained state and the constraint bound. The Lyapunov function used is a quadratic function of the sliding surface. In [12], a sliding mode control design with double power reaching law and variable power parameters is proposed. A changing magnitude of the parameters of the power terms of the reaching law is proposed, so as to improve the convergence rate of the approach phase. From the reaching law it follows that: (i) the control law involves a double power function of the sliding surface(s) with variable parameters; (ii) the gradient of the power function is higher when the sliding surface is farther from zero. As a result, the convergence speed of the sliding surface is faster for large values of the surface. Comparing the proposed SMC with the fast power reaching law, it achieves a faster convergence of the surface and a higher tracking precision in presence of time-varying disturbances. One deduces that this controller can be adapted to achieve output constraint requirements if the parameters of the double power function are chosen to fulfill the aforementioned FOC features.

In this work, a new robust controller is designed for the second-order plant model, considering asymmetric output constraints. The control design is based on dead-zone Lyapunov functions, aimed at achieving tracking error convergence. An improved output feedback term of the control law is proposed, consisting of the product between two user-defined functions of the tracking error: a basic increasing function and a sigmoid function that exhibits bounded but enhanced values for the tracking error equal to the constraint boundary. In this way, the output feedback term function is increasing with a steep section that leads to significantly higher but finite values for the tracking error equal to or higher than the constraint bound. The resulting control law yields an overlarge but bounded control effort for tracking error values equal to or higher than the prescribed constraint boundary. The main contributions over closely related output constraint control designs are:

- **Contribution Bi.** The output feedback term of the control law, and consequently, the control law, take on large but finite and user-defined values when the tracking error values are equal to or higher than the prescribed constraint bound. In contrast, in BLF-based output constraint controllers (for instance [3,4,13]) the output feedback term, and consequently, the control law, give infinite values when the tracking error is equal to the constraint bound, and it is not defined for tracking error values higher than the constraint bound.
- **Contribution Bii.** The output feedback term of the control law is equal to a user-defined output function when the tracking error is far from the constraint bound.

This implies that high-performance output feedback functions, for instance, the well-known power-law functions, can be used in the controller for these tracking error values. In contrast, in BLF-based output constraint controllers (see [3,4,13]), previously known user-defined functions cannot be used in the output feedback term for tracking error values far from the constraint boundary.

- **Contribution Biii.** The advantages of control design based on dead-zone Lyapunov functions are achieved, for instance, the absence of discontinuous signals in the controller whereas keeping robustness against disturbances or modeling error. In contrast, in BLF-based output constrain controllers (e.g., [4,13]), discontinuous signum type signals are used in the control law.

The system model, the reference model and the control goal are provided in Section 2, and the controller design and stability analysis in Section 3. The numerical simulation is provided in Section 4, and the conclusions are drawn in Section 5.

## 2. System Model, Reference Model and Control Goal

SISO second-order nonlinear model represent several systems, including mechatronic, biochemical and networked systems. This model usually includes disturbance terms caused by parametric uncertainty, modeling error, unmodeled parasitic dynamics and unknown external disturbances [14–16]. Consider a SISO second-order model with states  $x_1, x_2$ :

$$\frac{dx_1}{dt} = F_1 + F_{g1}x_2 + \delta_1, \quad (1a)$$

$$\frac{dx_2}{dt} = F_2 + b_m u + \delta_2. \quad (1b)$$

where  $x_1$  is the state to be controlled and  $u$  is the input. The model terms satisfy the following assumptions:

**Assumption 1.** The state  $x_2$  is bounded for  $u$  bounded,  $x_1 \in \mathbb{R}$ ,  $x_2 \in \mathbb{R}$ .

**Assumption 2.** The state  $x_1$  is measured and  $u$  is known.

**Assumption 3.**  $F_1, F_2$  and  $F_{g1}$  are known functions of  $x_1$  and  $x_2$ ; and  $b_m$  is known;  $\delta_1, \delta_2$  are unknown but bounded terms, with unknown bounds.

**Remark 1.** Assumption 1, that is,  $x_2$  is bounded if  $u$  bounded, is a case of input to state stability (ISS), which is commonly accomplished in practical scenario. Additionally, it implies that the nonlinear system can be globally stabilized in the presence of input saturation [17,18].

**Remark 2.** Assumption 2 implies that the measurement of  $x_1$  must be accurate for the application of the proposed controller. Hence, cases with significant noise are not allowed.

**Remark 3.** In Assumption 3, the bounded disturbance terms are caused by either parametric uncertainty, modeling error, unmodeled parasitic dynamics or unknown external disturbances [14–16]. Thus, these terms render the model more akin to practical scenario.

**Remark 4.** As the state  $x_2$  is not required to be known in the assumptions, real-time measurement of  $x_2$  is not needed. This unavailability of knowledge on some state is common in practical scenarios, due to either environmental disturbances, equipment cost or technical limitation [1].

The input  $u$  is constrained and its relationship with the unconstrained input signal  $u_{nc}$  is:

$$u = \begin{cases} u_{max} & \text{if } u_{nc} > u_{max} \\ u_{nc} & \text{if } u_{nc} \in [u_{min}, u_{max}] \\ u_{min} & \text{if } u_{nc} < u_{min} \end{cases} \quad (2)$$

where  $u_{min}$ ,  $u_{max}$  are constant input constraints.

**Remark 5.** The input saturation is due to limitations on physical structure of the actuator. Therefore, the input constraints  $u_{min}$ ,  $u_{max}$  are defined by operational limits [19].

An observer is considered in order to cope with the lack of knowledge on  $x_2$  and to avoid the risk of excessive increase of updated parameters, what is caused by input saturation (2). The observer model can be represented as:

$$\frac{d\hat{x}_1}{dt} = H_{ad1} + H_{g1} \hat{x}_2 \quad (3a)$$

$$\frac{d\hat{x}_2}{dt} = H_{ad2} + b_m u \quad (3b)$$

where  $\hat{x}_1$  is the estimate of  $x_1$ ;  $\hat{x}_2$  is the estimate of  $x_2$ ;  $H_{ad1}$ ,  $H_{g1}$ ,  $H_{ad2}$ ,  $b_m$  are known functions of  $x_1$  and  $x_2$ . The controller design in Section 3 uses this observer model instead of the system model (1).

**Remark 6.** Observers of this form are commonly used, for instance [20]. Also an example of observer for model (1) is provided in Appendix A, based on [21].

**Control goal.** Consider the tracking error  $e_1 = \hat{x}_1 - y_d$ , where  $y_d$  is desired output provided by the reference model

$$y_d = \frac{a_m}{p + a_m} \frac{a_m}{p + a_m} r \quad (4)$$

where  $r$  is the reference signal, and  $a_m$  is a user-defined positive constant, and  $p = d/dt$  is the differential operator [22,23]. The goal of the controller design is to formulate the control an update laws that achieve the convergence of the tracking error  $e_1$  towards the compact set  $\Omega_{e1} = \{e_1 : |e_1| \leq \varepsilon\}$ , and the constraint of the tracking error  $e_1$  within a region with upper bound  $\varepsilon_u$  or lower bound  $-\varepsilon_l$ , as user-defined, by using a controller with an output feedback function that has the following features: (i) it exhibits large but bounded and user-defined values for tracking error equal to or higher than the prescribed constraint bound; (ii) it involves a previously known user-defined function for tracking error values far from the constraint bound; (iii) it is non-decreasing with the tracking error, but its exact expression is user-defined. The width of the target compact set  $\varepsilon$ , the upper constraint bound  $\varepsilon_u$  and the lower constraint bound  $-\varepsilon_l$  are user-defined and constant, and  $\varepsilon$ ,  $\varepsilon_u$ ,  $\varepsilon_l$ , are positive.

**Remark 7.** The observer (3) is used in the controller design in order to cope with the lack of knowledge on the state  $x_2$  and the effect of input saturation (2). Thus, the observer model is used instead of the system model (1), and the estimates  $\hat{x}_1$ ,  $\hat{x}_2$  are used instead of  $x_1$ ,  $x_2$ .

### 3. Control Algorithm, Controller Design and Stability Analysis

#### 3.1. Control Algorithm

The controller is:

$$u_{nc} = \frac{1}{b_m} \left( -k_{c2} H_{g1}^2 e_2 - H_{ad2} - \frac{dT_{1f}}{dt} \right) \quad (5)$$

where

$$e_2 = \hat{x}_2 + T_{1f} \quad (6)$$

$T_{1f}$  is given by

$$\frac{dT_{1f}}{dt} = \frac{1}{\tau_{T1}} (-T_{1f} + T_1) \quad (7)$$

$$T_1 = \frac{(W - \dot{y}_d) + H_{ad1}}{H_{g1}} + k_f \text{sat}_{e1} \quad (8)$$

$$\text{sat}_{e1} = \begin{cases} \frac{e_1}{\varepsilon} \left( 2 - \frac{|e_1|}{\varepsilon} \right) & \text{for } e_1 \in (-\varepsilon, \varepsilon) \\ \text{sgn}(e_1) & \text{otherwise} \end{cases} \quad (9)$$

$$k_{c1}k_{c2} > \left( \frac{1}{4} \right); k_f \geq \mu_f + \varepsilon > 0 \quad (10)$$

where  $H_{ad1}$ ,  $H_{ad2}$ ,  $H_{g1}$  are functions of observer model (3), and: (i)  $\hat{x}_1$ ,  $\hat{x}_2$ , are state estimates provided by the observer; (ii)  $k_{c1}$ ,  $k_{c2}$  are user-defined positive constants; (iii)  $\tau_{T1}$  is the time constant of the signal  $T_{1f}$ ; (iv)  $k_f$  is the gain of the term for robustness against the error caused by the difference  $T_1 - T_{1f}$ .

In addition,

$$W = \phi_a f_{ehc} \quad (11)$$

where  $f_{ehc}$  is a function that leads to increased  $W$  value for  $e_1$  values equal to or higher than the constraint boundary:

$$\begin{aligned} f_{ehc} &\text{ is continuous and increasing with respect to } e_1; \\ f_{ehc} &\approx 1 \text{ for } e_1 \text{ values far from the bound but not exceeding it, with the bound being either } \varepsilon_u \text{ or } -\varepsilon_l; \\ f_{ehc} &\gg 1 \text{ for } e_1 \text{ values close to the bound } (\varepsilon_u \text{ or } -\varepsilon_l); \\ f_{ehc} &\gg 1 \text{ for } e_1 \text{ values exceeding the bound, that is, } e_1 > \varepsilon_u \text{ in case of upper bound } \varepsilon_u, \text{ or for } e_1 < -\varepsilon_l \text{ in case of lower bound } -\varepsilon_l. \end{aligned} \quad (12)$$

and  $\phi_a$  is a previously known user-defined function of  $f_{e1}$  that satisfies the properties:

$$\begin{aligned} \phi_a &= 0 \text{ for } f_{e1} = 0 \\ \phi_a &\neq 0 \text{ for } f_{e1} \neq 0 \\ \text{sgn}(\phi_a) &= \text{sgn}(f_{e1}) \neq 0 \text{ for } f_{e1} \neq 0 \\ \phi_a &\text{ is continuous with respect to } f_{e1} \\ \phi_a &\text{ is non-decreasing with } f_{e1} \end{aligned} \quad (13)$$

and

$$f_{e1} = \begin{cases} e_1 - \varepsilon & \text{for } e_1 \geq \varepsilon \\ 0 & \text{for } e_1 \in (-\varepsilon, \varepsilon) \\ e_1 + \varepsilon & \text{for } e_1 \leq -\varepsilon \end{cases} \quad (14)$$

The block diagram of the controller is shown in Figure 1.

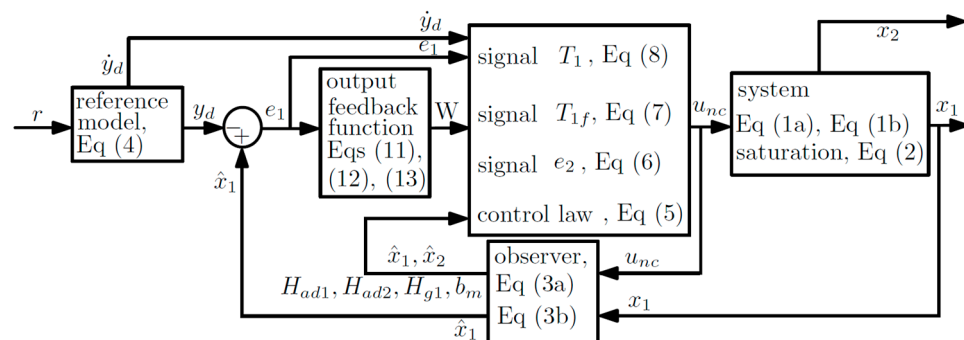


Figure 1. Block diagram of the proposed controller.

**Remark 8.** The output feedback function  $W$  (11) is a nonlinear function of the tracking error, and it exhibits the features mentioned in the control goal (Section 2):

- Bi: it takes on large but bounded and user-defined values for tracking error close, equal, or higher than the prescribed constraint boundary. It is achieved by using the sigmoid function  $f_{ehc}$ . This feature allows to achieve tracking error constraint.
- Bii: it involves a previously known user-defined function  $\phi_a$  for tracking error values far from the constraint bound, but it must satisfy conditions (13).

**Remark 9.** Functions  $\phi_a$  and  $f_{ehc}$  must satisfy conditions (13) and (12), respectively, but their exact expressions are user-defined.

**Remark 10.** An example of  $f_{ehc}$  for the case of upper constraint boundary  $\varepsilon_u$  and no lower constraint is:

$$f_{ehc} = (1 - f_T)(k_w - 1) + 1 \quad (15)$$

$$f_T = \frac{k_{1i}d_{ct}^n}{k_{oi} + k_{1i}d_{ct}^n}, \quad d_{ct} = \begin{cases} |\varepsilon_u - e_1| & \text{for } e_1 \leq \varepsilon_u \\ 0 & \text{for } e_1 > \varepsilon_u \end{cases}, \quad k_w > 1 \quad (16)$$

where  $k_w$  is a user-defined positive constant that specifies the enhancement of  $W$  for  $e_1$  values close to the bound  $\varepsilon_u$ , and  $k_{oi}$ ,  $k_{1i}$ ,  $n$  are user-defined positive constants. The main features of  $f_{ehc}$  (15) and  $f_T$  (16) are:

$$\begin{aligned} f_T &= 0 & \text{for } e_1 \geq \varepsilon_u \\ f_T &\approx 1 & \text{for } e_1 \ll \varepsilon_u \\ f_{ehc} &= k_w & \text{for } e_1 \geq \varepsilon_u \\ f_{ehc} &\approx 1 & \text{for } e_1 \ll \varepsilon_u \end{aligned}$$

so that  $f_{ehc}$  fulfills conditions (12).

**Remark 11.** An example of  $\phi_a$  is

$$\phi_a = k_{\phi 1}|f_{e1}|^\alpha \text{sgn}(f_{e1}) + k_{\phi 2}|f_{e1}|^3 \text{sgn}(f_{e1}) \quad (17)$$

which is based on the output feedback function used in [12,24]:

$$\phi_w = k_1|\bullet|^\alpha \text{sgn}(\bullet) + k_2|\bullet|^\beta \text{sgn}(\bullet) \quad (18)$$

where  $\alpha > 0$ ,  $\beta > 0$ ,  $k_1 > 0$ ,  $k_2 > 0$ , and the ranges and values  $\alpha \in [0.75, 1.6]$ ,  $\beta \in [0.75, 1.6]$ ,  $k_1 = 1.5$ ,  $k_2 = 0.8$  were used in [12], whereas  $\alpha = 5/8$ ,  $\beta = 3$ ,  $k_1 = 2.5 \times 42.4$ ,  $k_2 = 25 \times 42.4$ , were used in [24].

**Remark 12.** In BLF based output constraint control, the Lyapunov function of the type

$$V_{e1} = \frac{\varepsilon_u^\alpha}{\alpha} \frac{|e_1|^\alpha}{(\varepsilon_u - e_1)^\alpha}$$

with  $\alpha \geq 2$  is commonly used, being  $\varepsilon_u$  the upper constraint [3,4,13]. The gradient of  $V_{e1}$  is:

$$\frac{dV_{e1}}{de_1} = \varepsilon_u^{\alpha+1} \text{sgn}(e_1) \frac{|e_1|^{\alpha-1}}{(\varepsilon_u - e_1)^{\alpha+1}}$$

Therefore, the resulting output feedback function of the control law involves the term  $1/(\varepsilon_u - e_1)^\alpha$ , which leads to infinite control effort for  $e_1 = \varepsilon_u$ . A dead-zone modification can be incorporated in order to avoid discontinuous signals in the controller:

$$V_{e1} = \frac{\varepsilon_u^\alpha}{\alpha} \frac{|f_{e1}|^\alpha}{(\varepsilon_u - f_{e1})^\alpha}$$



$$\frac{dV_{e1}}{de_1} = \varepsilon_u^{\alpha+1} \operatorname{sgn}(f_{e1}) \frac{|f_{e1}|^{\alpha-1}}{(\varepsilon_u - f_{e1})^{\alpha+1}}$$

A BLF-based output constraint controller comprises Equations (5)–(10), (14) with the above gradient  $W = dV_{e1}/de_1$ , instead of the gradient (11)–(13). This allows to compare the proposed controller with respect to BLF-based output-constraint controller.

**Remark 13.** The developed controller can be applied to second-order systems subject to output constraint, for instance mechatronic, mechanical and biochemical systems [14–16]. Indeed, output constraint is common in practical scenarios, and it is caused by physical limitations [6,25].

### 3.2. Controller Design and Stability Analysis

**Theorem 1.** Consider the desired output  $y_d$  provided by the reference model (4), the second order model (1) subject to assumptions 1–3, the observer (3), the tracking error  $e_1 = \hat{x}_1 - y_d$ , which is constrained within a region with upper bound  $\varepsilon_u$  or lower bound  $-\varepsilon_l$ , as user-defined, the controller (5)–(14). Then, the tracking error  $e_1$  converges asymptotically to  $\Omega_{e1} = \{e_1 : |e_1| \leq \varepsilon\}$ , where the bound  $\varepsilon$  is user-defined, positive and constant; and  $e_1$  is constrained within a region with upper bound or lower bound  $-\varepsilon_l$ , as user-defined.

**Proof.** **Task 1 [Definition of the tracking error  $e_1$  and its time derivative].** Recall the observer (3). We define the tracking error as  $e_1 = \hat{x}_1 - y_d$ . Differentiating with respect to time, yields

$$\frac{de_1}{dt} = \frac{d\hat{x}_1}{dt} - \frac{dy_d}{dt}$$

Incorporating the  $d\hat{x}_1/dt$  expression (3a), yields:

$$\frac{de_1}{dt} = H_{g1}\hat{x}_2 + H_{ad1} - \dot{y}_d \quad (19)$$

**Task 2 [Definition of ( $V_{e1}$ ) the dead-zone subsystem Lyapunov function for the tracking error, its gradient, and its properties].** The control goal includes the achievement of output tracking and the formulation of a controller with an output feedback function satisfying the stated features, being the exact output feedback function defined by the user. Therefore,  $V_{e1}$ , the Lyapunov function of the tracking error  $e_1$ , is defined as an integral of its gradient  $W$  and the  $V_{e1}$  and  $W$  properties are defined, but the exact  $W$  expression is not. Thus, the properties of  $V_{e1}$ , are set as:

$$\begin{aligned} V_{e1} &= 0 \quad \text{for } f_{e1} = 0 \\ V_{e1} &> 0 \quad \text{for } f_{e1} \neq 0 \\ V_{e1} &\text{ is continuous with respect to } f_{e1}, \\ V_{e1} &\text{ is bounded for } f_{e1} \text{ bounded} \\ V_{e1} &\text{ is non-decreasing with } |f_{e1}| \end{aligned} \quad (20)$$

where

$$f_{e1} = \begin{cases} e_1 - \varepsilon & \text{for } e_1 \geq \varepsilon \\ 0 & \text{for } e_1 \in (-\varepsilon, \varepsilon) \\ e_1 + \varepsilon & \text{for } e_1 \leq -\varepsilon \end{cases} \quad (21)$$

The main properties of  $f_{e1}$  (21) are:

$$\begin{aligned} f_{e1} &= 0 \quad \text{for } e_1 \in [-\varepsilon, \varepsilon] \\ f_{e1} &\neq 0 \quad \text{for } e_1 \notin [-\varepsilon, \varepsilon] \\ \operatorname{sgn}(f_{e1}) &= \operatorname{sgn}(e_1) \neq 0 \quad \text{for } e_1 \notin [-\varepsilon, \varepsilon] \\ f_{e1} &\in \mathcal{L}_\infty \quad \text{for } e_1 \in \mathcal{L}_\infty \end{aligned}$$

Thus, the  $V_{e1}$  properties (20) in terms of  $f_{e1}$  lead to the properties in terms of  $e_1$ :

$$\begin{aligned} V_{e1} &= 0 \text{ for } e_1 \in [-\varepsilon, \varepsilon] \\ V_{e1} &> 0 \text{ for } e_1 \notin [-\varepsilon, \varepsilon] \\ V_{e1} &\text{ is continuous with respect to } e_1, \text{ and it is bounded for } e_1 \text{ bounded} \\ V_{e1} &\text{ is non-decreasing with increasing } |e_1| \end{aligned}$$

In order to obtain convergence of  $e_1$  to the compact set  $\Omega_{e1}$  and to facilitate its proof, we require the gradient

$$W \triangleq \frac{dV_{e1}}{df_{e1}} \quad (22)$$

to fulfill the following properties:

$$\begin{aligned} W &= 0 \text{ for } f_{e1} = 0 \\ W &\neq 0 \text{ for } f_{e1} \neq 0 \\ \text{sgn}(W) &= \text{sgn}(f_{e1}) \neq 0 \text{ for } f_{e1} \neq 0 \\ W &\text{ is continuous with respect to } f_{e1} \\ W &\text{ is non-decreasing with } f_{e1} \end{aligned} \quad (23)$$

As a consequence, the properties of  $W$  in terms of  $e_1$  are:

$$\begin{aligned} W &= 0 \text{ for } e_1 \in [-\varepsilon, \varepsilon] \\ W &\neq 0 \text{ for } e_1 \notin [-\varepsilon, \varepsilon] \\ \text{sgn}(W) &= \text{sgn}(e_1) = \text{sgn}(f_{e1}) \neq 0 \text{ for } e_1 \notin [-\varepsilon, \varepsilon] \\ W &\text{ is continuous with respect to } e_1 \\ W &\text{ is non-decreasing with } e_1 \end{aligned} \quad (24)$$

A practical way to define  $V_{e1}$  and  $W$  is to define a  $W$  function of  $f_{e1}$  that satisfies properties (23), and then to determine  $V_{e1}$  from

$$V_{e1} = \int_0^{f_{e1}} W df_{e1}$$

What leads to fulfillment of  $V_{e1}$  properties (20). The integral form of  $V_{e1}$  for the case of asymmetrical is considered in Appendix B. The basic choice of  $W$  is  $W = k_{c1}f_{e1}$  which fulfills properties (23) and leads to  $V_{e1} = (1/2)k_{c1}f_{e1}^2$ , which fulfills properties (20).

**Task 3 [Determination of  $dV_{e1}/dt$  and incorporation of the output feedback term].** The time derivative of  $V_{e1}$  can be expressed as

$$\frac{dV_{e1}}{dt} = \frac{dV_{e1}}{de_1} \frac{de_1}{dt} \quad (25)$$

From the definition of  $W$  (22), the definition of  $f_{e1}$  (21) and the properties of  $W$  (24), it follows that

$$\frac{dV_{e1}}{de_1} = W$$

Combining with Equation (25), yields

$$\frac{dV_{e1}}{dt} = W \frac{de_1}{dt}$$

Incorporating the  $de_1/dt$  expression (19), yields

$$\frac{dV_{e1}}{dt} = -W^2 + W(H_{g1}\hat{x}_2 + (W - \dot{y}_d) + H_{ad1}) \quad (26)$$



where the term  $W^2$  has been subtracted and added to provide the non-positive term  $-W^2$  in the  $dV_{e1}/dt$  expression, what allows obtaining convergence of  $e_1$  towards the expected compact set.

**Task 4 [Definition of the second backstepping state  $e_2$ ].** Equation (26) can be rewritten as

$$\frac{dV_{e1}}{dt} \leq -W^2 - k_f H_{g1} |W| + W H_{g1} \left( \hat{x}_2 + \frac{(W - \dot{y}_d) + H_{ad1}}{H_{g1}} + k_f \text{sign}(W) \right) \quad (27)$$

where the term  $-k_f H_{g1} |W|$  has been added and subtracted to provide robustness against an error term that will arise later because of the filtering approximation. To avoid the discontinuous signal  $\text{sign}(f_{e1})$ , we notice from the  $W$  properties (24) that:

$$|W| = W \text{sat}_{e1}$$

with

$$\text{sat}_{e1} = \begin{cases} \frac{e_1}{\varepsilon} \left( 2 - \frac{|e_1|}{\varepsilon} \right) & \text{for } e_1 \in (-\varepsilon, \varepsilon) \\ \text{sgn}(e_1) & \text{otherwise} \end{cases}$$

$$\frac{d\text{sat}_{e1}}{de_1} = \begin{cases} \frac{2}{\varepsilon} \left( 1 - \frac{|e_1|}{\varepsilon} \right) & \text{for } e_1 \in (-\varepsilon, \varepsilon) \\ 0 & \text{otherwise} \end{cases}$$

Therefore, Equation (27) can be rewritten as

$$\frac{dV_{e1}}{dt} \leq -W^2 - k_f H_{g1} |W| + W H_{g1} \left( \hat{x}_2 + \frac{(W - \dot{y}_d) + H_{ad1}}{H_{g1}} + k_f \text{sat}_{e1} \right)$$

Or equivalently,

$$\frac{dV_{e1}}{dt} \leq -W^2 - k_f H_{g1} |W| + W H_{g1} (\hat{x}_2 + T_1) \quad (28)$$

where

$$T_1 = \frac{(W - \dot{y}_d) + H_{ad1}}{H_{g1}} + k_f \text{sat}_{e1}$$

If  $e_2$  were defined as  $e_2 = \hat{x}_2 + T_1$ , the term  $dT_1/dt$  in its time derivative would involve an undesired ‘explosion of terms’. To avoid this effect, the DSC strategy involves the use of a filtered signal in the definition of  $e_2$  [26,27]. Therefore, we propose a definition of  $e_2$  with a filtered  $T_1$ , denoted as  $T_{1f}$ . Equation (28) can be rewritten as

$$\frac{dV_{e1}}{dt} \leq -W^2 - k_f H_{g1} |W| + W H_{g1} (e_2 + \delta_{t1f}) \quad (29)$$

where

$$e_2 = \hat{x}_2 + T_{1f} \quad (30)$$

and  $T_{1f}$  is given by

$$\frac{dT_{1f}}{dt} = \frac{1}{\tau_{T1}} (-T_{1f} + T_1) \quad (31)$$

and  $\delta_{t1f} = T_1 - T_{1f}$  is the error caused by filtering. In the current DSC strategy, it is assumed that the time derivative of the input signal of the filter is bounded, see [26]. In the case of filter given in Equation (31), that assumption would be  $dT_1/dt \in L_\infty$ . Equation (31) and  $T_1 \in \mathcal{L}_\infty$  imply  $T_{1f} \in \mathcal{L}_\infty$  and, consequently,  $\delta_{t1f} \in \mathcal{L}_\infty$ , so that  $\delta_{t1f} \leq \mu_f$ , where  $\mu_f$  is a positive constant. Therefore, Equation (29) can be rewritten as:

$$\frac{dV_{e1}}{dt} \leq -k_f H_{g1} |W| + \mu_f |H_{g1}| |W| - W^2 + W H_{g1} e_2 \quad (32)$$

If  $k_f$  is chosen such that  $k_f \geq \mu_f > 0$ , then  $-k_f|H_{g1}||W| + \mu_f|H_{g1}||W| \leq 0$ , so that Equation (32) leads to

$$\frac{dV_{e1}}{dt} \leq -W^2 + WH_{g1}e_2$$

**Task 5 [Determination of the time derivative  $de_2/dt$ ].** Differentiating  $e_2$  (Equation (30)) with respect to time, yields

$$\frac{de_2}{dt} = \frac{d\hat{x}_2}{dt} + \frac{dT_{1f}}{dt}$$

Substituting expression for  $d\hat{x}_2/dt$  (Equation (3b)), yields

$$\frac{de_2}{dt} = H_{ad2} + b_mu + \frac{dT_{1f}}{dt} \quad (33)$$

where  $dT_{1f}/dt$  is given by Equation (31).

**Task 6 [Definition of  $V_{e2}$ , the dead-zone Lyapunov function for  $e_2$ , determination of its time derivative and formulation of the control law].** As we need to obtain the convergence of the state  $e_2$  (30) to  $\Omega_{e2} = \{e_2 : |e_2| \leq \varepsilon\}$ , we choose  $V_{e2}$ , the dead-zone Lyapunov-like function for the state  $e_2$  as

$$V_{e2} = \frac{1}{2}f_{e2}^2 \quad (34)$$

where

$$f_{e2} = \begin{cases} e_2 - \varepsilon & \text{for } e_2 \geq \varepsilon \\ 0 & \text{for } e_2 \in (-\varepsilon, \varepsilon) \\ e_2 + \varepsilon & \text{for } e_2 \leq -\varepsilon \end{cases} \quad (35)$$

The main properties of  $V_{e2}$  are:

$$\begin{aligned} V_{e2} &= 0 \quad \text{for } e_2 \in [-\varepsilon, \varepsilon] \\ V_{e2} &> 0 \quad \text{for } e_2 \notin [-\varepsilon, \varepsilon] \\ V_{e2} &\text{ is continuous with respect to } e_2, \\ V_{e2} &\text{ is bounded for } e_2 \text{ bounded} \end{aligned}$$

Differentiating  $V_{e2}$  (34) with respect to time, yields

$$\frac{dV_{e2}}{dt} = f_{e2} \frac{de_2}{dt}$$

Incorporating the  $de_2/dt$  expression (Equation (33)), and arranging, yields

$$\frac{dV_{e2}}{dt} = -k_{c2} H_{g1}^2 f_{e2} e_2 + f_{e2} \left( k_{c2} H_{g1}^2 e_2 + H_{ad2} + b_mu + \frac{dT_{1f}}{dt} \right) \quad (36)$$

To counteract the effect of the term  $f_{e2} (k_{c2} H_{g1}^2 e_2 + H_{ad2} + dT_{1f}/dt)$  we choose the control law for  $u_{nc}$  as:

$$u_{nc} = \frac{1}{b_m} \left( -k_{c2} H_{g1}^2 e_2 - H_{ad2} - \frac{dT_{1f}}{dt} \right) \quad (37)$$

Considering moments of no input saturation ( $u = u_{nc}$ ), substituting (37) into Equation (36), yields

$$\frac{dV_{e2}}{dt} = -k_{c2} H_{g1}^2 f_{e2} e_2 \quad (38)$$

**Task 7 [Determination of the  $dV_{e1}/dt + dV_{e2}/dt$  expression under the formulated controller and arrangement in terms of a non-positive function of  $e_1$ ].** Adding expressions for  $dV_{e1}/dt$  (32) and  $dV_{e2}/dt$  (38), yields:

$$\frac{dV_{e1}}{dt} + \frac{dV_{e2}}{dt} \leq -k_f H_{g1} |W| + \mu_f |H_{g1}| |W| - k_{c1} (\beta_1 + \beta_2) W^2 + W H_{g1} e_2 - k_{c2} H_{g1}^2 f_{e2} e_2 \quad (39)$$

where  $\beta_1, \beta_2$  are positive constants that satisfy  $\beta_1 + \beta_2 = 1$ ,  $\beta_1 \in (0, 1)$ ,  $\beta_2 \in (0, 1)$ , and the term  $W H_{g1} e_2$  exhibits no non-positive nature, so that it must be counteracted by the non-positive terms

$$-k_f |H_{g1}| |W| - k_{c1} (\beta_1 + \beta_2) W^2 - k_{c2} H_{g1}^2 f_{e2} e_2$$

To this end,  $e_2$  is expressed in terms of  $f_{e2}$  and the term  $-k_f |H_{g1}| |W| + \mu_f |H_{g1}| |W| + W H_{g1} e_2$  appearing in Equation (39) is arranged at what follows. From the definition of  $f_{e2}$  (35) it follows that

$$f_{e2} = e_2 + d_{e2}.$$

$$d_{e2} = \begin{cases} -\varepsilon & \text{for } e_2 \geq \varepsilon \\ -e_2 & \text{for } e_2 \in (-\varepsilon, \varepsilon) \\ \varepsilon & \text{for } e_2 \leq -\varepsilon \end{cases}$$

Therefore,

$$e_2 = f_{e2} - d_{e2}, \quad |d_{e2}| \leq \varepsilon$$

In view of this expression, the term  $W H_{g1} e_2$  can be rewritten as  $W H_{g1} e_2 = W H_{g1} f_{e2} + W H_{g1} (-d_{e2})$ . Hence,  $W H_{g1} e_2 \leq W H_{g1} f_{e2} + \varepsilon |W| |H_{g1}|$ . Therefore, the term  $-k_f |H_{g1}| |W| + \mu_f |H_{g1}| |W| + W H_{g1} e_2$  leads to

$$-k_f |H_{g1}| |W| + \mu_f |H_{g1}| |W| + W H_{g1} e_2 \leq -(k_f - \mu_f - \varepsilon) |H_{g1}| |W| + H_{g1} W f_{e2} \quad (40)$$

If  $k_f$  is chosen such that  $k_f \geq \mu_f + \varepsilon \geq \varepsilon > 0$ , then  $-(k_f - \mu_f - \varepsilon) |H_{g1}| |W| \leq 0$ , and, consequently, Equation (40) leads to  $-k_f |H_{g1}| |W| + \mu_f |H_{g1}| |W| + W H_{g1} e_2 \leq H_{g1} W f_{e2}$ . Combining this expression with Equation (41) leads to

$$\frac{dV_{e1}}{dt} + \frac{dV_{e2}}{dt} \leq -k_{c1} (\beta_1 + \beta_2) W^2 + H_{g1} W f_{e2} - k_{c2} H_{g1}^2 f_{e2} e_2 \quad (41)$$

Some properties of  $f_{e2}$  (35) are:

$$\begin{aligned} \operatorname{sgn}(f_{e2}) &= \operatorname{sgn}(e_2) \neq 0 \text{ for } e_2 \notin [-\varepsilon, \varepsilon]; \\ |f_{e2}| &< |e_2| \text{ for } e_2 \notin [-\varepsilon, \varepsilon]; f_{e2} = 0 \text{ for } e_2 \in [-\varepsilon, \varepsilon] \end{aligned}$$

Hence,

$$\begin{aligned} e_2 f_{e2} &= |e_2| |f_{e2}| > f_{e2}^2 \text{ for } e_2 \notin [-\varepsilon, \varepsilon] \\ e_2 f_{e2} &= 0 = f_{e2}^2 \text{ for } e_2 \in [-\varepsilon, \varepsilon] \end{aligned}$$

Combining these properties, yields  $e_2 f_{e2} \geq f_{e2}^2$ . Therefore,  $-k_{c2} H_{g1}^2 f_{e2} e_2 \leq -k_{c2} H_{g1}^2 f_{e2}^2$ . Substituting this into Equation (41), yields

$$\frac{dV_{e1}}{dt} + \frac{dV_{e2}}{dt} \leq -k_{c1} (\beta_1 + \beta_2) W^2 + H_{g1} W f_{e2} - k_{c2} H_{g1}^2 f_{e2}^2 \quad (42)$$

where  $\beta_1$  and  $\beta_2$  are user-defined positive constants that satisfy  $\beta_1 + \beta_2 = 1$ ,  $\beta_1 \in (0, 1)$ ,  $\beta_2 \in (0, 1)$ . Equation (42) can be arranged as

$$\begin{aligned} & \frac{dV_{e1}}{dt} + \frac{dV_{e2}}{dt} \leq -k_{c1}\beta_1 W^2 \\ & + (-1)k_{c1}\beta_2 \left[ W^2 + 2W \left( \frac{1}{2k_{c1}\beta_2} H_{g1} f_{e2} \right) + 4k_{c1}k_{c2}\beta_2 \left( \frac{H_{g1}f_{e2}}{2k_{c1}\beta_2} \right)^2 \right] \end{aligned} \quad (43)$$

If  $k_{c1}$ ,  $k_{c2}$  values are chosen such that  $k_{c1}k_{c2} \geq (1/4)\beta_2^{-1}$ , then  $4k_{c1}k_{c2}\beta_2 \geq 1$  and Equation (43) lead to

$$\begin{aligned} \frac{dV_{e1}}{dt} + \frac{dV_{e2}}{dt} & \leq -k_{c1}\beta_1 W^2 + (-1)k_{c1}\beta_2 \left[ W + \frac{1}{2k_{c1}\beta_2} H_{g1} f_{e2} \right]^2 \\ & \leq -k_{c1}\beta_1 W^2 \leq 0 \end{aligned} \quad (44)$$

**Task 8. [Integration of the  $d(V_{e1} + V_{e2})V/dt$  expression, and determination of the convergence of  $e_1$ ].** Arranging and integrating Equation (44) yields

$$V_{e1} + V_{e2} + k_{c1}\beta_1 \int_{t_0}^t W^2 dt \leq V_{e1|t_0} + V_{e2|t_0}$$

where  $V_{e1|t_0}$ ,  $V_{e2|t_0}$  are  $V_{e1}$  (20),  $V_{e2}$  (34) at time  $t_0$ . From the above expression it follows that  $V_{e1} \leq V_{e1|t_0} + V_{e2|t_0}$ ;  $V_{e2} \leq V_{e1|t_0} + V_{e2|t_0}$ ;  $k_{c1}\beta_1 \int_{t_0}^t W^2 dt \leq V_{e1|t_0} + V_{e2|t_0}$ . Therefore, applying Barbalat's lemma [28], we obtain the result that  $f_{e1}$  converges asymptotically to zero, and the definition of  $f_{e1}$  (21) implies that  $e_1 = \hat{x}_1 - y_d$  converges asymptotically to  $\Omega_{e1}$ ,  $\Omega_{e1} = \{e_1 : |e_1| \leq \varepsilon\}$ , for moments of no input saturation. This completes the proof.

**Task 9 [Definition of the structure of the output feedback term of the controller].** From the control law (37) it follows that the output feedback function  $W$  has a straightforward effect on the control law. To achieve the output feedback function features stated in the control goal, we propose the output feedback function structure

$$W = \phi_a f_{ehc} \quad (45)$$

where  $f_{ehc}$  is a function that increases the value of  $W$  for  $e_1$  values equal to or higher than the constraint bound ( $\varepsilon_u$  or  $-\varepsilon_l$ ):

it is continuous and increasing with respect to  $e_1$ ;

$f_{ehc} \approx 1$  for  $e_1$  values far from the bound, either  $\varepsilon_u$  or  $-\varepsilon_l$ ;

$f_{ehc} \gg 1$  for  $e_1$  values close to the bound ( $\varepsilon_u$  or  $-\varepsilon_l$ );

$f_{ehc} \gg 1$  for  $e_1 > \varepsilon_u$  in case of upper bound  $\varepsilon_u$ , or  $e_1 < -\varepsilon_l$  in case of lower bound  $-\varepsilon_l$ .

The function  $\phi_a$  is a previously known user-defined function of  $f_{e1}$  that satisfies the  $W$  properties (23):

$$\begin{aligned} \phi_a &= 0 \quad \text{for } f_{e1} = 0 \\ \phi_a &\neq 0 \quad \text{for } f_{e1} \neq 0 \\ \text{sgn}(\phi_a) &= \text{sgn}(f_{e1}) \neq 0 \quad \text{for } f_{e1} \neq 0 \\ \phi_a &\text{ is continuous with respect to } f_{e1} \\ \phi_a &\text{ is non-decreasing with } f_{e1} \end{aligned} \quad (46)$$

Some examples of  $\phi_a$  and  $f_{ehc}$  are given in Equations (15), (17) and (18).  $\square$

### 3.3. Discussion of Results

The developed controller design considers a second-order model with model uncertainties and output error constraints. Asymmetric tracking error constraint is considered. The main closed loop features are: (i) the tracking error converges to a compact set whose width is user-defined so that it depends on neither model terms, modeling error, nor model coefficients; (ii) output error constraint is achieved by using enhanced control effort; (iii) dis-

continuous signals are avoided in the controller. In addition, the output feedback term of the control law takes on large values for tracking error values close to the constraint boundary.

The overall steps of the control design are: (i) definition of the tracking error  $e_1$  and its time derivative; (ii) definition of the dead-zone Lyapunov-like function for the tracking error ( $V_{e1}$ ), its gradient and its properties; (iii) determination of the time derivative  $dV_{e1}/dt$ , and incorporation of the output feedback term; (iv) definition of the second backstepping state  $e_2$ ; (v) determination of the time derivative  $de_2/dt$ ; (vi) definition of the dead-zone Lyapunov-like function for  $e_2$  ( $V_{e2}$ ), determination of its time derivative, and formulation of the control law; (vii) determination of the  $d(V_{e1} + V_{e2})/dt$  expression under the formulated controller; (viii) integration of the  $d(V_{e1} + V_{e2})/dt$  expression and determination of the boundedness of the closed-loop signals and convergence of  $e_1$ ; (ix) definition of the structure of the output feedback term of the controller.

The control design procedure aims at rendering the time derivative of the overall Lyapunov function non-positive. The backstepping strategy is used as the basic framework for the control design, and the filtering approximation of the DSC strategy is used in order to avoid an ‘explosion of complexity’. However, several improvements are incorporated in the basic backstepping-DSC procedure to achieve contributions Bi, Bii and Biii: an additional robustness term is used in order to tackle the effect of the approximation error; dead-zone modification is used in the Lyapunov-like function; a new output feedback stabilizing term is incorporated in the time derivative  $dV_{e1}/dt$ .

The dead-zone modification of the Lyapunov functions has the following advantages: it allows for avoiding discontinuous signals in the controller signals, while robustness to disturbances or uncertainty terms is ensured; it facilitates guaranteeing the convergence of the tracking error to the user-defined region. Dead-zone Lyapunov functions have been mainly applied to robust control design: early studies are presented in [22,29,30], whereas recent studies in [31–35]. In turn, the theory of the dead-zone Lyapunov function keeps several of the advantages of the current Lyapunov function, including its usefulness for proving global asymptotic stability and for formulating nonlinear control which can be noticed in [36,37].

The output feedback function  $W$  (45) is a nonlinear function of the tracking error, and it exhibits the advantages mentioned in the control goal (Section 2):

- Bi: it takes on large but bounded and user-defined values for tracking error close, equal, or higher than the prescribed constraint bound, which is done through the sigmoid function  $f_{ehc}$ . This function satisfies conditions (12), whereas its exact expression and the level of enhancement  $k_w$  are user-defined. The tracking error constraint is achieved through this control effort enhancement.
- Bii: it involves a previously known user-defined output feedback function  $\phi_a$  for tracking error values far from the constraint bound. This function satisfies conditions (46), whereas its exact expression is user-defined so that high-performance functions are allowed, for instance the double power law (18).

#### 4. Numerical Simulation

In this section, the controller and convergence results stated in Theorem 1 are illustrated through simulation for a continuous bioreactor, using the input constraint condition (2), the reference model (4) and the controller Equations (5)–(14). We consider a continuous culture of *Gluconacetobacter diazotrophicus* with constant culture volume and substrate sensor dynamics, being the measured substrate concentration ( $s_s$ ) the output to be controlled and the dilution rate ( $D$ ) the control input. Its model comprises biomass and substrate mass balance equations [20] and substrate sensor dynamics [38]:

$$\frac{ds_s}{dt} = -\frac{1}{\tau_1}s_s + \frac{1}{\tau_1}s \quad (46a)$$

$$\frac{ds}{dt} = (s_{in} - s)D - y_s\mu x \quad (46b)$$

$$\frac{dx}{dt} = (\mu - D)x \quad (46c)$$

where  $x$  is the biomass concentration;  $x_2 = s$  is the substrate concentration;  $s_{in}$  is the inlet substrate concentration;  $\mu$  is the specific growth rate,  $y_s$  is the yield coefficient;  $y_s\mu$  is the substrate uptake rate,  $D = Q/V$  is the dilution rate;  $Q$  is the feeding flow rate;  $V$  is the liquid volume [20] and  $x_1 = s_s$  is the output of the substrate sensor [38]. The equations for substrate measurement (46a) and substrate concentration (46b) correspond to model (1) and fulfill assumptions 1 to 3. The bacterium strain, initial medium composition, temperature, pH, dissolved oxygen concentration and mixing conditions correspond to the batch cultivation presented in [39]. The specific growth rate expression is

$$\mu = \mu_{max} \left( 1 - \frac{x}{x_{max}} \right)^f$$

The parameters of substrate and biomass models (46b), (46c) are fitted to batch culture data presented in [39], giving as result:  $\mu_{max} = 0.01484 \text{ h}^{-1}$ ;  $X_{max} = 0.32 \text{ g/L}$ ;  $f = 1.607$ ;  $y_s = 0.0234$ ;  $m_s = 0.22425$ . In addition,  $s_{in} = 180 \text{ g/L}$  and  $\tau_1 = 0.03 \text{ h}$  were used for the continuous plant models (46). The width of the tracking error convergence set  $\Omega_{e1}$ , the constraint boundary and the input bound are set to  $\varepsilon = 1.3$ ;  $\varepsilon_u = 1.9$ ;  $D_{max} = 4$ . The observer used is based on that of [20], and the state estimation results in  $\hat{x}_1 \approx x_1$ , and  $\hat{x}_2 \approx x_2$ .

The parameters of control Equations (5)–(11) are chosen as  $a_m = 0.1$ ;  $\tau_{T1} = 0.01$ ;  $k_f = 0.195$ . We use an asymmetric form of (17) for  $\phi_a$ :

$$\phi_a = \begin{cases} k_{\phi 1} |f_{e1}|^\alpha \operatorname{sgn}(f_{e1}) + k_{\phi 2} |f_{e1}|^3 \operatorname{sgn}(f_{e1}) & \text{for } e_1 > 0 \\ -k_{c1} f_{e1} & \text{for } e_1 \leq 0 \end{cases}$$

with  $k_{c1} = 0.5$ ;  $k_{\phi 1} = 0.0002164$ ;  $k_{\phi 2} = 4.408$ ;  $\alpha = 1.417$ , so that the  $W$  function is:

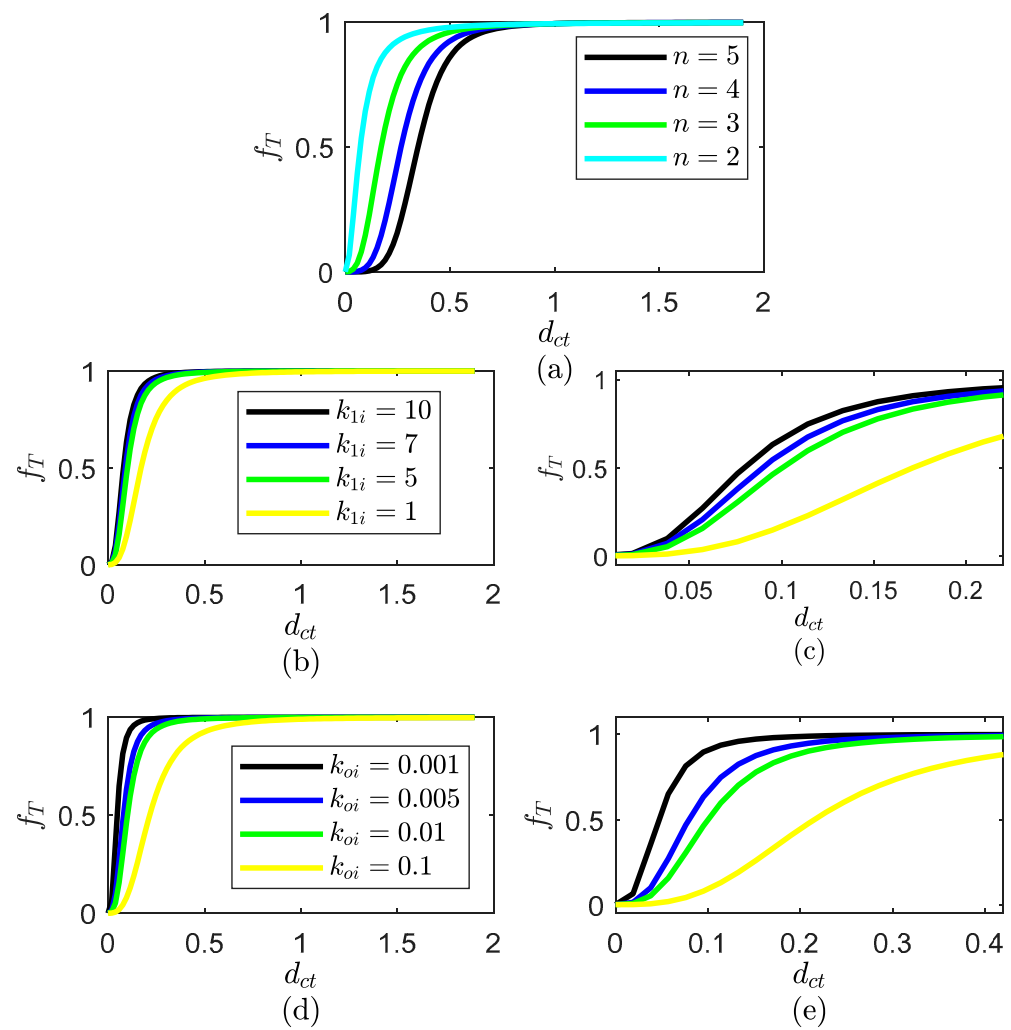
$$W = \begin{cases} \phi_a f_{ehc} & \text{for } e_1 > 0 \\ -k_{c1} f_{e1} & \text{for } e_1 \leq 0 \end{cases}$$

The function (15) is used for  $f_{ehc}$ : the effect of  $k_{oi}$ ,  $k_{1i}$ ,  $n$  on the  $f_T$  curve is shown in Figure 2, whereas their effect on the  $f_{ehc}$  and  $W$  curves is shown in Figure 3. The  $f_T$  curve (Figure 2) shows that: the  $f_T$  function is continuous and smooth;  $f_T \approx 0$  for low  $d_{ct}$  values;  $f_T \approx 1$  for high values. In addition, steeper curve is obtained with high  $n$ , high  $k_{1i}$ , low  $k_{oi}$ .

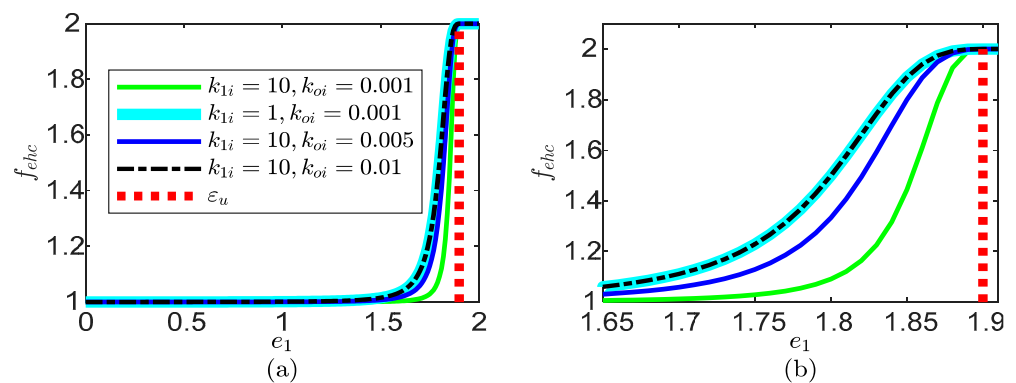
The  $f_{ehc}$  curve (Figure 3a,b) shows that: the  $f_{ehc}$  function is continuous and smooth;  $f_{ehc} \approx 1$  for low  $e_1$  values;  $f_{ehc} \approx k_w$  for  $e_1$  close to  $\varepsilon_u$ . In addition, steeper form is obtained with high  $n$ , high  $k_{1i}$ , low  $k_{oi}$ .

The  $W$  curve (Figure 3c,d) shows that: the  $W$  function is continuous and smooth;  $W \approx \phi_a$  for low  $e_1$  values;  $W \approx k_w \phi_a$  for  $e_1$  close to  $\varepsilon_u$ . In summary, the definition of  $f_{ehc}$  parameters to obtain the desired  $W$  features is simple, while involving the user-defined dependence for low  $e_1$  values and for  $e_1$  close to  $\varepsilon_u$ .

The values of  $f_{ehc}$  are chosen to be  $k_{oi} = 0.001$ ,  $10$ ,  $n = 3$ . The simulation is performed using MATLAB version R2014 with ode45 function. The trajectories of the tracking error  $e_1$ , the backstepping state  $e_2$ , the control input  $D$ , the output  $x_1 = S$  and the desired output  $y_d$  for these  $n$ ,  $k_{oi}$ ,  $k_{1i}$  values and different values of  $e_{t0}$  are shown in Figures 4 and 5.

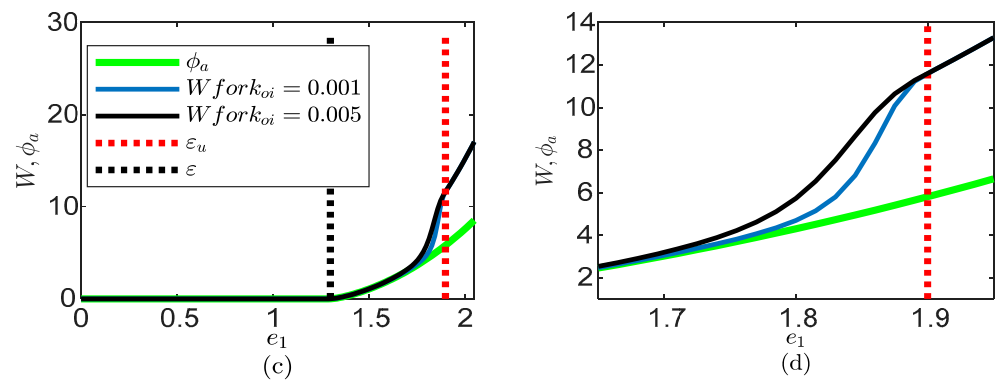


**Figure 2.** Effect of  $d_{ct}$  on the  $f_T$  function (16), for different values of  $n, k_{oi}, k_{1i}$ : (a) curve of  $f_T$  for  $k_{oi} = 0.005, k_{1i} = 1$ ; (b) curve of  $f_T$  for  $k_{oi} = 0.005$  and  $n = 3$ ; (c) detail of  $f_T$  curve for  $k_{oi} = 0.005$  and  $n = 3$ ; (d) curve of  $f_T$  for  $k_{1i} = 10$  and  $n = 3$ ; (e) detail of  $f_T$  curve for  $k_{1i} = 10$  and  $n = 3$ .

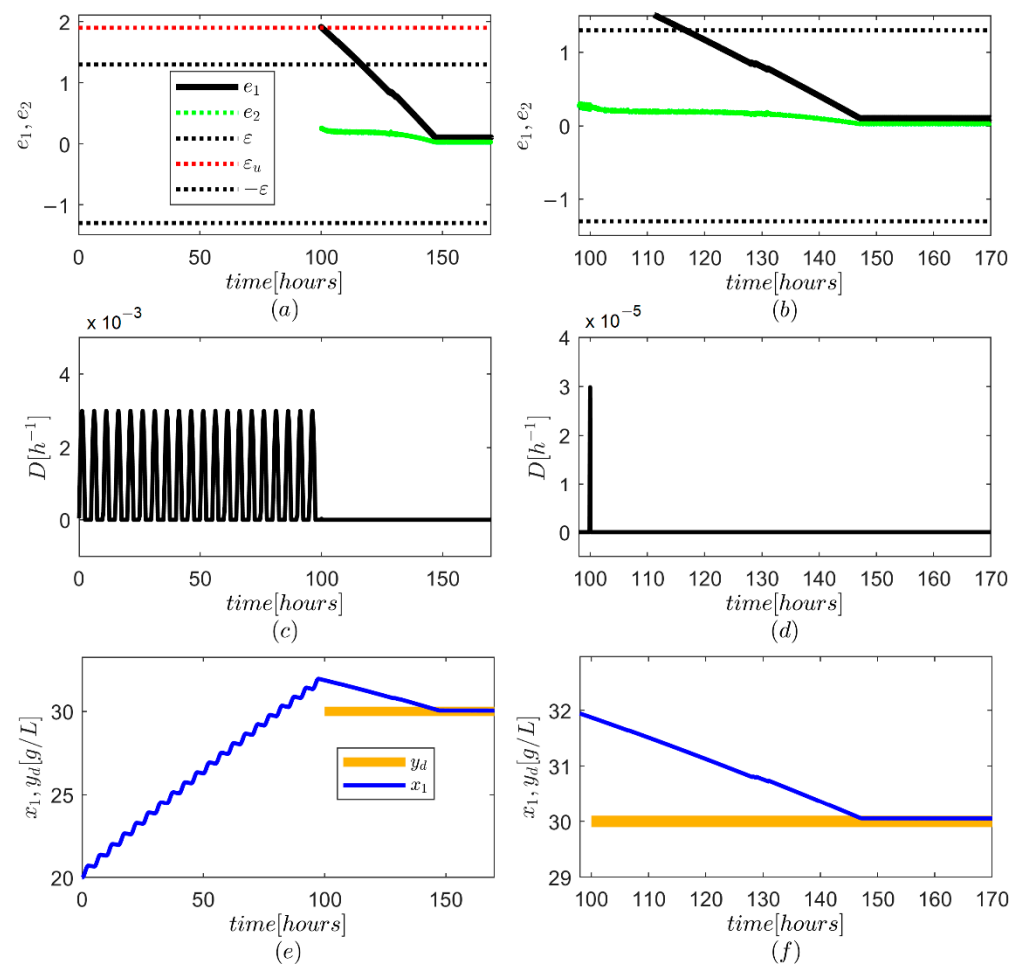


**Figure 3.** Cont.

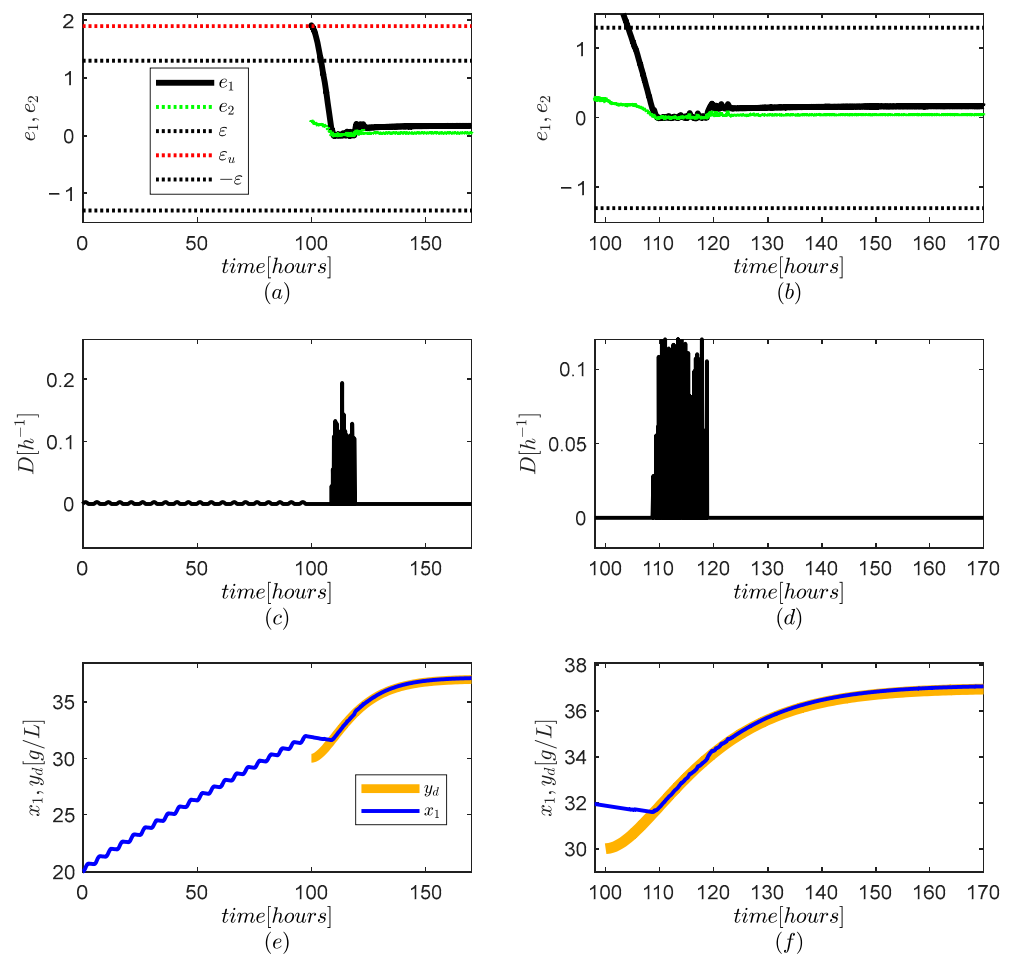




**Figure 3.** Effect of the tracking error  $e_1$  on the  $f_{ehc}$  function (15) and  $W = \phi_a f_{ehc}$  for  $\varepsilon_u = 1.9$ : (a) curve of  $f_{ehc}$  for  $k_w = 2$ ,  $n = 3$ ; (b) detail of the  $f_{ehc}$  curve; (c) curve of  $W$  and  $\phi_a$  for  $n = 3$ ,  $k_{1i} = 10$ ,  $k_{oi} = 0.001$ , and  $k_{oi} = 0.005$ ; (d) detail of the  $W$  and  $\phi_a$  curves.



**Figure 4.** Simulation results for  $r = 30$ : (a) Trajectory of the tracking error  $e_1$  and backstepping state  $e_2$ ; (b) detail of the trajectories of  $e_1$  and  $e_2$ ; (c) trajectory of the control input  $D$ ; (d) detail of the trajectory of the control input; (e) trajectory of the output  $x_1$  and desired output  $y_d$ ; (f) detail of the trajectories of  $x_1, y_d$ .



**Figure 5.** Simulation results for  $r = 37$ : (a) Trajectory of the tracking error  $e_1$  and backstepping state  $e_2$ ; (b) detail of the trajectories of  $e_1$  and  $e_2$ ; (c) trajectory of the control input  $D$ ; (d) detail of the trajectory of the control input; (e) trajectory of the output  $x_1$  and desired output  $y_d$ ; (f) detail of the trajectories of  $x_1, y_d$ .

Simulations (Figures 4 and 5) verify that:

- the developed controller achieves a fast asymptotic convergence of the tracking error  $e_1$  to the compact set  $\Omega_{e1} = \{e_1 : |e_1| \leq \varepsilon\}$ .
- the dependence of the control effort on the tracking error is remarkably enhanced with the applied  $f_{ehc}$  function (Figures 3 and 4), thus avoiding violation of output constraint. This is mostly noticed at initial time when the tracking error is high and outside  $\Omega_{e1}$ .

## 5. Conclusions

In this paper, we proposed a robust controller for systems described by the second-order input-output model, considering asymmetric tracking error constraints. The main contribution over closely related studies is that the tracking error convergence is achieved by using a control law whose output feedback term is user-defined and bounded, comprising large but finite user-defined values for tracking error values equal to or higher than the constraint boundary, and a user-defined function for tracking error values far from the constraint boundary. This is a significant contribution that remedies two important limitations of common output constraint control designs: the infinite control effort for tracking error equal to or higher than the constraint boundary, and the impossibility of using previously known user-defined functions in the output feedback function for tracking error values far from the constraint boundary. As another contribution, the control design is based on

the dead-zone Lyapunov function, which facilitates the achievement of convergence to a compact set with user-defined size, avoidance of discontinuous signals in the controller and robustness to model uncertainty or disturbances.

To achieve the aforementioned features, the main improvements made in the control design procedure are: (i) dead zone modification is used in the Lyapunov function; (ii) the Lyapunov function for the tracking error is defined as an integral in terms of its gradient; (iii) the non-positive stabilizing term in the time derivative of the overall Lyapunov function is chosen to be the square of the output feedback term; (iv) the output feedback term is defined as the product between a sigmoid function and an increasing function, both of them depending on the tracking error.

The main limitations of the proposed controller are:

- The design is only valid for the second-order systems, so that systems of higher-order are not allowed.
- Finite-time stabilization is not guaranteed. That is, the time for convergence of the tracking error  $e_1$  to its compact set  $\Omega_{e1}$  is not straightforwardly defined by the user. Indeed, some trial-and-error effort is needed for achieving a desired convergence time.
- The control gain  $b_m$  is considered completely known.

The simulations show several of the advantages mentioned in the contributions including (i) convergence of the tracking error to a compact set whose width is user-defined; (ii) the tracking error is constrained by using the proposed output feedback term in the control law; (iii) the input signal is continuous.

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## Appendix A

The observer of [21], using the terms of model (1) is:

$$\begin{aligned}\frac{d\hat{x}_1}{dt} &= -|F_{g1}| \left( \omega \bar{x}_1 + \left( k + \frac{1}{4\omega} \right) \psi_{x1} + sat_{x1} \hat{\theta} \right) + F_1 + F_{g1} \hat{x}_2 \\ \frac{d\hat{x}_2}{dt} &= -F_{g1} \omega \left( \left( k + \frac{1}{4\omega} \right) \psi_{x1} + sat_{x1} \hat{\theta} \right) + F_2 + b_m u \\ \frac{d\hat{\theta}}{dt} &= \gamma |F_{g1}| |\psi_{x1}|\end{aligned}$$

where

$$\psi_{x1} = \begin{cases} \bar{x}_1 - \varepsilon & \text{for } \bar{x}_1 \geq \varepsilon \\ 0 & \text{for } \bar{x}_1 \in [-\varepsilon, \varepsilon] \\ \bar{x}_1 + \varepsilon & \text{for } \bar{x}_1 \leq -\varepsilon \end{cases}$$

$$\text{sat}_{x1} = \begin{cases} 1 & \text{for } \bar{x}_1 \geq \varepsilon \\ \frac{1}{\varepsilon} \bar{x}_1 & \text{for } \bar{x}_1 \in [-\varepsilon, \varepsilon] \\ -1 & \text{for } \bar{x}_1 \leq -\varepsilon \end{cases}$$

$$\bar{x}_1 = \hat{x}_1 - x_1,$$

$\varepsilon$  is the width of the convergence region of  $\bar{x}_1$ , and  $\omega, k, \gamma, \varepsilon$ , are user-defined positive constants. Therefore, the terms of observer form (3) are:

$$H_{ad1} = |F_{g1}| \left( \omega \bar{x}_1 + \left( k + \frac{1}{4\omega} \right) \psi_{x1} + \text{sat}_{x1} \hat{\theta} \right) - F_1$$

$$H_{g1} = F_{g1}$$

$$H_{ad2} = -F_{g1} \omega \left( \left( k + \frac{1}{4\omega} \right) \psi_{x1} + \text{sat}_{x1} \hat{\theta} \right) + F_2$$

## Appendix B

The function  $W$  can be asymmetrical:

$$W = \begin{cases} W_u & \text{for } f_{e1} \geq 0 \\ W_l & \text{for } f_{e1} < 0 \end{cases}$$

In this case,  $V_{e1}$  is determined from

$$V_{e1} = \begin{cases} \int_0^{f_{e1}} W_u df_{e1} & \text{for } f_{e1} \geq 0 \\ \int_0^{f_{e1}} W_l df_{e1} & \text{for } f_{e1} < 0 \end{cases}$$

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