

## RESEARCH ARTICLE

# Control Synthesis of Nonholonomic Mobile Robots Under Time-Varying Delays and Input Saturation: Experimental Validation

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**ABSTRACT** This paper presents a new control synthesis methodology for nonholonomic mobile robots subjected to time-varying delays and input saturation constraints. The proposed control method is based on smooth static nonlinear functions, leading to a simpler structure than other available control strategies for this class of systems. In addition, the convergence to a certain target position with guaranteed exponential decay rate can be proved for any orientation error. To this end, a nonlinear Lyapunov-Krasovskii functional has specifically been designed to deal with the inherent discontinuity of the kinematics model as well as the presence of time-varying delays. Thus, the control design can efficiently be addressed by means of Linear Matrix Inequalities (LMIs). Finally, the effectiveness of the proposed control design algorithm is validated through simulation and experimentally using a two-wheeled mobile robot.

**INDEX TERMS** Time delay systems, control synthesis, nonholonomic mobile robot, Lyapunov-Krasovskii functional, linear matrix inequalities.

## I. INTRODUCTION

The control synthesis applied to nonholonomic robotic systems is a challenging problem due to their complex nature. Indeed, this class of systems are not controllable (linearly) around the equilibria [1]. Hence, smooth-feedback stability conditions for any initial orientation error cannot be ensured, as discussed in [2]. This fact motivated the design of discontinuous control strategies, such as sliding mode control [3], [4], [5], backstepping control [6] or adaptive control [7], [8], among others. However, these control methods may lead to undesirable phenomena, such as chattering effects and high frequency oscillations as well as high complexity. To circumvent such drawbacks, other proposals resorted to smooth control laws under Takagi-Sugeno fuzzy models and parallel distributed compensation (PDC) in [9]. In this case, the discontinuities and nonlinear terms in the kinematics model were treated by means of sector

nonlinearity approach [10] in order to find global stabilizing conditions based on LMIs. However, the main drawback is that only local stability can be proved around the zero-orientation error.

It is noteworthy that many control applications involving nonholonomic mobile robots are subjected to time delays [1] and input saturation constraints [11]. The presence of delays or input saturation in the control loop may lead to poor performance or instability if they are not taken into account in control synthesis [12]. The infinite-dimensional nature of time delay systems prevents from the use of classical techniques to deal with stability analysis. Instead, the Lyapunov-Krasovskii (LK) method [13], [14] allows to prove the stability analysis of time delay systems by checking a *finite* number of Linear Matrix Inequalities (LMIs) [15], which can efficiently be solved using commercially available semidefinite programming tools, such as SeDuMi [16] or LMI Toolbox [17] at the expense of extra conservatism in the worst-case delay estimation. Motivated by this, many efforts have been addressed to reduce such conservatism,

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ranging from the construction of more sophisticated LK functionals [18] and the improvement of the integral bounding techniques based on the Jensen’s inequality [19] to other more advanced bounding approaches, such as Wirtinger’s inequality [20], auxiliary based-functions inequalities [21] or Bessel-Legendre inequalities [22], together with The Reciprocally Convex Lemma [23] to deal with time-varying delays. Other extensions of these methods to nonlinear systems subjected to time-varying delays and input saturation constraints can be found in [24] with nonlinear functions satisfying Lipschitz conditions.

The Lyapunov-Krasovskii method [1], [25], [26] has also been applied for stabilization of nonholonomic systems subjected to time delays. Nevertheless, the latter control strategies involve high nonlinear dynamic equations which are difficult to apply in practice. Considering that the control code programmed at the mobile robot should be kept as simple as possible because of the hardware capabilities limitations [27], the research of easy-to-implement control algorithms is of practical interest [28]. Moreover, in spite of the practical importance of considering input saturation constraints in control synthesis of mobile robots owing to the limited capacities of actuators [11], [29], [30], to the best of the authors’ knowledge there are no previous results that simultaneously consider input saturation and time-varying delays for nonholonomic systems.

Motivated by this, the objective of this paper is three-fold: (i) design a smooth control strategy for nonholonomic robots subjected to time-varying delays and input saturation constraints, (ii) give a theoretical proof of the exponential convergence to the target position for any orientation error, and (iii) provide experimental results to support the validity of the theoretical conclusions. Hence, the main contributions can be summarized as follows:

- The proposed control law resorts to smooth nonlinear static functions, which is of simpler structure and less complexity than other available strategies, such as those based on sliding mode control [3], [4], [5], backstepping control [6] or adaptive control [7], [8].
- The exponential convergence with guaranteed decay rate under time-varying delays and input saturation constraints is proved via Lyapunov-Krasovskii and LMIs for any initial orientation error.
- The effectiveness of the proposed control synthesis is experimentally validated in a two-wheeled mobile robot in the sense to reduce to the greatest extent the time required to reach the target position (settling time) for a given worst-case delay.

The rest of the paper is organized as follows: Section II describes the problem statement and the proposed control strategy. Section III presents the main results consisting of the LMIs for global exponential stability analysis. Section IV depicts some simulation results, which are experimentally validated in Section V. Finally, some conclusions are gathered in Section VI.

## II. PROBLEM STATEMENT

Consider the following kinematic model for a nonholonomic mobile robot (see Fig. 1):

$$\begin{aligned} \dot{x} &= \text{sat}(v^{(h)}) \cos(\theta), \\ \dot{y} &= \text{sat}(v^{(h)}) \sin(\theta), \\ \dot{\theta} &= \text{sat}(w^{(h)}) \end{aligned} \tag{1}$$

where

$$\text{sat}(v^{(h)}) = \begin{cases} -\bar{v} & \text{if } v^{(h)} \leq -\bar{v} \\ v^{(h)} & \text{if } -\bar{v} < v^{(h)} < \bar{v} \\ \bar{v} & \text{if } v^{(h)} \geq \bar{v} \end{cases},$$

$$\text{sat}(w^{(h)}) = \begin{cases} -\bar{w} & \text{if } w^{(h)} \leq -\bar{w} \\ w^{(h)} & \text{if } -\bar{w} < w^{(h)} < \bar{w} \\ \bar{w} & \text{if } w^{(h)} \geq \bar{w} \end{cases} \tag{2}$$

being  $v^{(h)} \equiv v(t - h(t))$  and  $w^{(h)} \equiv w(t - h(t))$  the control actions corresponding to the linear and angular velocities with the saturation constraints  $\bar{v}$  and  $\bar{w}$  respectively. Both control actions are subjected to a time-varying input delay  $h(t)$  satisfying  $h_1 \leq h(t) \leq h_2$  for some scalars  $0 \leq h_1 \leq h_2$ . Let  $(x_r, y_r)$  be a prescribed reference 2D position to be reached, and:

$$\begin{aligned} \rho &= \sqrt{(x - x_r)^2 + (y - y_r)^2}, \\ \phi &= \text{atan2}(y - y_r, x - x_r). \end{aligned} \tag{3}$$

*Remark 1:* The function  $\text{atan2} : \mathcal{R}^2 \rightarrow (-\pi, \pi)$  is equivalent to a four-quadrant arctangent function [31] defined as:

$$\begin{aligned} &\text{atan2}(y, x) \\ &= \begin{cases} 0 & \text{if } (x, y) = (0, 0) \\ \arctg\left(\frac{y}{x}\right) + \frac{\pi}{2} \text{sign}(y) (1 - \text{sign}(x)) & \text{otherwise} \end{cases} \end{aligned} \tag{4}$$

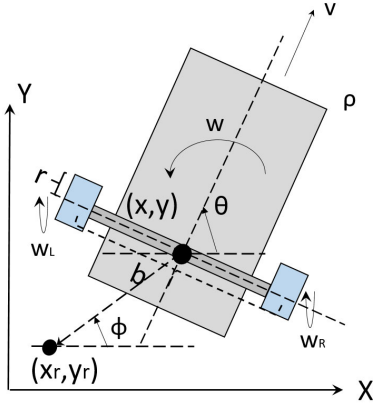
where

$$\text{sign}(a) = \begin{cases} 1 & \text{if } a \geq 0 \\ 0 & \text{otherwise} \end{cases} \tag{5}$$

Applying the above definitions, system (1) can be expressed in polar coordinates as:

$$\begin{aligned} \dot{\rho} &= \text{sat}(v^{(h)}) \cos(\phi - \theta), \\ \dot{\phi} &= -\frac{\text{sat}(v^{(h)})}{\rho} \sin(\phi - \theta), \\ \dot{\theta} &= \text{sat}(w^{(h)}) \end{aligned} \tag{6}$$

Taking into account the existing trade-off between settling time and tolerance against delays, the objective is to design the control laws for  $v$  and  $w$  in such a way that the mobile robot can reach the reference position  $(x_r, y_r)$  in a certain time interval, while enlarging the worst-case delay as far as possible without reaching the input saturation levels  $\bar{v}$  and  $\bar{w}$


**FIGURE 1.** Nonholonomic mobile robot: control position system.

respectively. For this purpose, let us consider the following control laws for  $v$  and  $w$ :

$$\begin{aligned} v &= -K_v \rho \cos(\phi - \theta), \\ w &= -K_w \sin(\phi - \theta) \end{aligned} \quad (7)$$

where  $K_v, K_w$  are control gains to be designed.

### III. STABILITY ANALYSIS

This section addresses the stability analysis of the closed-loop system formed by (6) and the control law (7). Before proceeding, let us present the following definition, useful later for Theorem 1:

**Definition 1** ([32]): The position error  $(x - x_r, y - y_r)$  in (1) is globally exponentially stable with decay rate  $\alpha$  in the sense of the orientation error  $\phi - \theta$  if there exists a constant  $F(t_0)$  for any  $t_0 \geq 0$  such that  $\rho \equiv \rho(t) \leq F(t_0)\rho(t_0)e^{-\alpha(t-t_0)}, \forall t \geq t_0$ , regardless of the values of the angles  $\phi, \theta$  given in (3).

The fulfilment of the LMI conditions given in Theorem 1 below proves the exponential convergence of the mobile robot (6) with the control (7) to the target position  $(x_r, y_r)$  for any time-varying delay  $h_1 \leq h(t) \leq h_2$  and any orientation error  $\phi - \theta$  without exceeding the input saturation level:

**Theorem 1:** Given  $K_v, K_w < \bar{w}$ ,  $h_1, h_2$ , and  $\alpha$ , the closed-loop system formed by (1) and (7) is globally exponentially stable in the sense of Definition 1 satisfying  $v < \bar{v}$  and  $w < \bar{w}$  for any initial distance to the target position  $\rho_0 < \bar{\rho}_0$  if there exist scalars  $\gamma_1, \gamma_2, \gamma_3$ , symmetric matrices  $Q_1, Q_2, R_1, R_4, R_6 \in \mathcal{R}^2, X_1, X_2 \in \mathcal{R}^4, Z_1, Z_2 \in \mathcal{R}^2 > 0$ , and matrices  $Y_1, Y_2 \in \mathcal{R}^4, R_2, R_3, R_5 \in \mathcal{R}^2 > 0$  satisfying the following LMIs  $\forall [i, j, k, m, n, p] = [1, 2] \times [1, 2] \times [1, 2] \times [1, 2] \times [1, 2] \times [1, 2]$ :

$$\begin{aligned} \hat{\Xi}_{ij} + R_1 + h_1^2 R_4 + \tau^2 R_6 + h_1 (R_2 + R_2^T) + \tau (R_3 + R_3^T) \\ + h_1 \tau (R_5 + R_5^T) + h_1 Q_1 + \tau Q_2 < \frac{1}{\rho_0^2} I, \end{aligned}$$

$$\begin{bmatrix} \hat{\Xi}_{ij} & \bar{K}_v^T \\ \bar{K}_v & \bar{v}^2 \end{bmatrix} > 0,$$

$$\hat{\Xi}_{ij} > 0,$$

$$\begin{aligned} \begin{bmatrix} \hat{\Pi}_{ijkmp} & \hat{\Pi}_{2, kp}^T \bar{Z} \\ (*) & -\bar{Z} \end{bmatrix} < 0, \\ \begin{bmatrix} \bar{Z}_2 & 0 \\ 0 & \bar{Z}_2 \end{bmatrix} - (1 - \hat{\delta}_{5n}) \begin{bmatrix} X_1 & Y_1 \\ (*) & 0 \end{bmatrix} - \hat{\delta}_{5n} \begin{bmatrix} 0 & Y_2 \\ (*) & X_2 \end{bmatrix} \geq 0, \end{aligned} \quad (8)$$

where

$$\begin{aligned} \Xi_{ij} &= \text{diag}(\hat{\Xi}_{ij}, 0, 0) + \frac{1}{h_1} \text{diag}(0, Q_1, 0) \\ &\quad + \frac{1}{\tau} \text{diag}(0, 0, Q_2) + \bar{R}, \\ \hat{\Xi}_{11} &= I_2 \otimes (\gamma_1 + \gamma_2 + \gamma_3), \\ \hat{\Xi}_{12} &= I_2 \otimes (\gamma_1 + \gamma_2 - \gamma_3), \\ \hat{\Xi}_{21} &= I_2 \otimes (\gamma_1 - \gamma_2 + \gamma_3), \\ \hat{\Xi}_{22} &= I_2 \otimes (\gamma_1 - \gamma_2 - \gamma_3), \\ \hat{\Pi}_{ijkmp} &= E_1^T \hat{\Pi}_{0, ij} E_1 + He(E_1^T \hat{\Pi}_{1, ijmp} E_2) \\ &\quad + He(\hat{\Pi}_{4, kp}^T \bar{R} \hat{\Pi}_{3, n}) + E_1^T Q_1 E_1 - E_3^T Q_1 E_3 \\ &\quad + E_1^T Q_2 E_1 - E_4^T Q_2 E_4 \\ &\quad - \mathcal{D}_1^T \mathcal{W}^T \bar{Z}_1 \mathcal{W} \mathcal{D}_1 - \mathcal{D}_2^T \bar{\mathcal{W}}^T \hat{\mathcal{T}}_n \bar{\mathcal{W}} \mathcal{D}_2, \\ \hat{\Pi}_{2, kp} &= \hat{\Pi}_{21, k} E_1 + \hat{\Pi}_{22, p} E_2, \\ \bar{Z}_1 &= \begin{bmatrix} Z_1 & 0 \\ 0 & 3Z_1 \end{bmatrix}, \bar{Z}_2 = \begin{bmatrix} Z_2 & 0 \\ 0 & 3Z_2 \end{bmatrix}, \\ \bar{Z} &= h_1^2 Z_1 + \tau^2 Z_2, \hat{\mathcal{T}}_n = \begin{bmatrix} \bar{Z}_2 + \hat{\mathcal{T}}_{1, n} & \hat{\mathcal{T}}_{2, n} \\ (*) & \bar{Z}_2 + \hat{\mathcal{T}}_{3, n} \end{bmatrix}, \\ \hat{\mathcal{T}}_{1, n} &= \hat{\delta}_{5n} X_1, \hat{\mathcal{T}}_{3, n} = (1 - \hat{\delta}_{5n}) X_2, \\ \hat{\mathcal{T}}_{2, n} &= (1 - \hat{\delta}_{5n}) Y_1 + \hat{\delta}_{5n} Y_2, \\ \mathcal{W} &= \begin{bmatrix} 1 & -1 & 0 \\ 1 & 1 & -2 \end{bmatrix} \otimes I_2, \\ \bar{\mathcal{W}} &= I_2 \otimes \mathcal{W}, \quad \tau = h_2 - h_1. \end{aligned} \quad (9)$$

being  $He(\cdot) = (\cdot) + (\cdot)^T$  for any matrix  $(\cdot)$ , and

$$\begin{aligned} \hat{\Pi}_{0, ij} &= \begin{bmatrix} \hat{\Omega}_{0, ij} & 0 \\ (*) & \hat{\Omega}_{0, ij} \end{bmatrix}, \hat{\Pi}_{1, ijmp} = \begin{bmatrix} \hat{\Omega}_{1, ijp} & \hat{\Omega}_{2, mp} \\ \hat{\Omega}_{3, ijp} & \hat{\Omega}_{4, mp} \end{bmatrix}, \\ \hat{\Pi}_{21, k} &= \begin{bmatrix} \alpha & \hat{\Omega}_{5, k} \\ 0 & \hat{\Omega}_{6, k} \end{bmatrix}, \hat{\Pi}_{22, p} = \begin{bmatrix} -K_v \hat{\delta}_{6p} & 0 \\ 0 & 0 \end{bmatrix}, \\ \hat{\Pi}_{3, n} &= [E_1^T \quad h_1 E_5^T \quad \tau(1 - \hat{\delta}_{5n}) E_6^T + \tau \hat{\delta}_{5n} E_7^T]^T, \\ \hat{\Pi}_{4, kp} &= [\hat{\Pi}_{2, kp}^T \quad (E_1 - E_3)^T \quad (E_3 - E_4)^T]^T, \\ \hat{\Omega}_{0, ij} &= \alpha (\gamma_1 + \gamma_2 \delta_{1i} + \gamma_3 \delta_{2j}), \\ \hat{\Omega}_{1, ijp} &= -\frac{1}{2} K_v \hat{\delta}_{6p} (\gamma_1 + \gamma_2 \delta_{1i} + \gamma_3 \delta_{2j}), \\ \hat{\Omega}_{2, mp} &= -\frac{1}{4} K_w \gamma_3 \hat{\delta}_{6p} \hat{\delta}_{4m}, \quad i = 1, 2, \\ \hat{\Omega}_{3, ijp} &= -\frac{1}{4} K_v (\gamma_2 \delta_{2j} + \gamma_3 \delta_{1i}) \hat{\delta}_{6p}, \\ \hat{\Omega}_{4, mp} &= -\frac{1}{4} K_w \gamma_2 \hat{\delta}_{6p} \hat{\delta}_{4m}, \\ \hat{\Omega}_{5, k} &= K_w \hat{\delta}_{3k}, \quad \hat{\Omega}_{6, k} = \alpha - K_w \hat{\delta}_{3k}. \\ \bar{K}_v &= [K_v \ 0 \ 0 \ 0 \ 0 \ 0], \end{aligned}$$

$$\bar{R} = \begin{bmatrix} R_1 & R_2 & R_3 \\ R_1^T & R_4 & R_5 \\ R_3^T & R_5^T & R_6 \end{bmatrix},$$

$$E_i = e_i \otimes I_2, \quad i = 1, 2, \dots, 7,$$

$$e_1 = [1, 0, 0, 0, 0, 0, 0],$$

$$e_2 = [0, 1, 0, 0, 0, 0, 0],$$

$$e_3 = [0, 0, 1, 0, 0, 0, 0],$$

$$e_4 = [0, 0, 0, 1, 0, 0, 0],$$

$$e_5 = [0, 0, 0, 0, 1, 0, 0],$$

$$e_6 = [0, 0, 0, 0, 0, 1, 0],$$

$$e_7 = [0, 0, 0, 0, 0, 0, 1],$$

$$D_1 = [E_1^T \quad E_3^T \quad E_5^T]^T,$$

$$D_2 = [E_3^T \quad E_2^T \quad E_6^T \quad E_2^T \quad E_4^T \quad E_7^T]^T,$$

$$\hat{\delta}_{11} = \hat{\delta}_{21} = \hat{\delta}_{31} = -1,$$

$$\hat{\delta}_{12} = \hat{\delta}_{22} = \hat{\delta}_{32} = 1,$$

$$\hat{\delta}_{41} = 1, \quad \hat{\delta}_{42} = e^{-\alpha h_2} - 1,$$

$$\hat{\delta}_{51} = 0, \quad \hat{\delta}_{52} = 1,$$

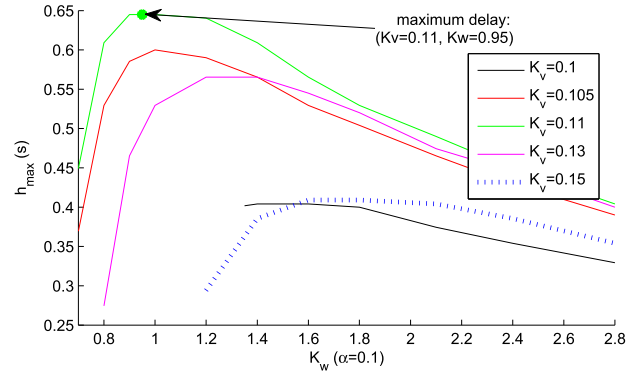
$$\hat{\delta}_{61} = e^{\alpha h_1}, \quad \hat{\delta}_{62} = e^{\alpha h_2}. \tag{10}$$

*Proof:* See Appendix VI-A. □

*Remark 2:* From  $\xi_1, \xi_2$  in (14), notice that  $\xi_1 = 0, \xi_2 = 0$  if and only if  $\rho = 0$  for any orientation error  $\phi - \theta$ . Noting that  $\dot{V} < 0$  (43) has a quadratic form with the augmented state vector  $\bar{\xi}$  in (40), one can deduce that the fulfilment of the LMI conditions of Theorem 1 implies the global exponential convergence of the position error in the sense of Definition 1. The proposed LKF (Lyapunov-Functional Candidate)  $V$  given in (12) has been crucial to this end, allowing to give LMI conditions of Theorem 1 that prove global exponential stability with the proposed smooth control function (7) in spite of the discontinuity of the nonholonomic kinematics model (6). In addition, the structure of  $V$  allows to take advantage of the Wirtinger’s inequality [20] to reduce the conservatism of the integral inequalities in (27), and the Reciprocally Convex Lemma [23] to deal with time-varying delays in (30).

*Remark 3:* Note that Theorem 1 cannot be directly used for control design since the controller gains  $K_v, K_w$  must be set in advance. From the fact that small values for  $K_v$  and  $K_w$  usually lead to slow convergence, and high values for  $K_v, K_w$  tend to closed-loop instability in the presence of delays or input saturation, it can be seen that there exists an optimal choice for such gains in the sense of maximum worst-case delay whose values can be found by dichotomic search with the aid of Theorem 1, starting from sufficiently small gains.

*Remark 4:* Note that the above conditions do not require to bound the time-derivative of  $\tau(t)$ . Hence, arbitrarily fast time-varying delays are allowed. This feature allows to include sampling-based sensors with zero-order hold mechanisms, leading to time-varying delay functions of sawtooth structure with positive unitary slope and discontinuities at each sampling instant. For the case of time-varying delays



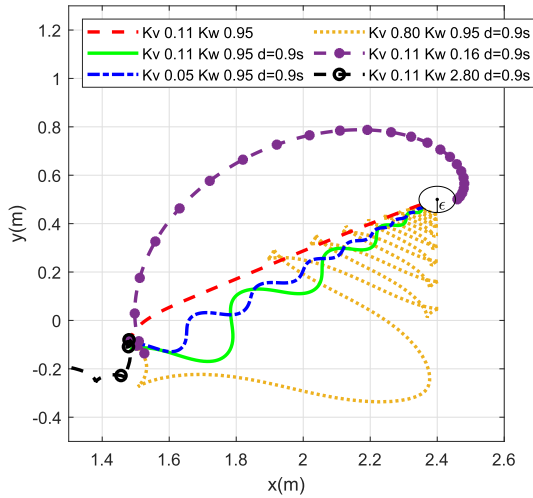
**FIGURE 2.** Worst-case delay  $h_{max} = h_1 = h_2$  as a function of the controller gains  $K_v$  and  $K_w$  obtained by Theorem 1 for a decay rate  $\alpha = 0.1$  and maximum linear and angular velocities of  $0.19m/s$  and  $2.82rad/s$  respectively.

with known bounded derivative, a new component of the form  $\int_{t-h(t)}^{t-h_1} \xi^T(s)Q\xi(s)ds$  for a certain symmetric matrix  $Q > 0$  can be included in the LKF proposed in (12) to reduce conservatism. Since the experimental platform includes the effect of sampling period, we have intentionally addressed the case of time-varying delays with unknown time-derivative.

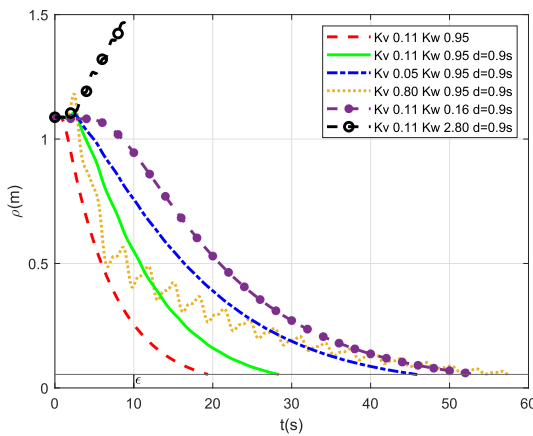
#### IV. SIMULATION EXAMPLES

Consider a nonholonomic robot with a kinematic model described by (1) and the proposed control law (7). The goal is to design the control gains  $K_v, K_w$  in order to reach the target position with prescribed maximum settling time around 30s with error less than 98% (decay rate  $\alpha = 0.1$ ), where the worst-case delay  $h_2$  is enlarged to the greatest extent. Applying Theorem 1 with  $\alpha = 0.1$  and  $h_1 = h_2$  for different values of  $K_v$ , the worst-case delay  $h_2$  is depicted in Fig 2 as a function of  $K_w$ . It can be seen that the maximum worst-case delay  $h_2$  is obtained for  $K_v = 0.11$  and  $K_w = 0.95$  (green solid line in Fig 2).

In order to give a comparison between the control performance with the designed values  $K_v = 0.11$  and  $K_w = 0.95$  and other choices for  $K_v, K_w$ , different simulations have been carried out by means of MATLAB-Simulink and the Simscape Multibody library. A detailed description of the simulation platform can be found in [33]. The obtained results have been depicted in Fig. 3 and Fig. 4. Fig. 3 shows the trajectories in the XY plane described by the mobile robot, and Fig. 4 represents the time evolution of the distance  $\rho$  to the target position. The first simulation (red dashed line) has been performed without delays considering the designed values  $K_v = 0.11$  and  $K_w = 0.95$  (nominal case). Then, the worst-case delay has been increased until  $h_2 = 0.9s$ . It can be appreciated that the settling time of 30s is maintained (green line). The rest of simulations have been carried out choosing other different values for  $K_v, K_w$  with  $h_2 = 0.9s$ . Comparatively, the control gains  $K_v = 0.11$  and  $K_w = 0.95$  give the fastest response in comparison to other choices: smaller values for the controller gains lead to slower convergence (for instance  $K_v = 0.05, K_w = 0.95$ , blue dashed line), and higher values for the controller gains lead



**FIGURE 3.** Simulation data: comparison of the system trajectories in the XY plane for different control gains  $K_V$ ,  $K_W$  with the designed values  $K_V = 0.11$  and  $K_W = 0.95$  in the presence of time delays.

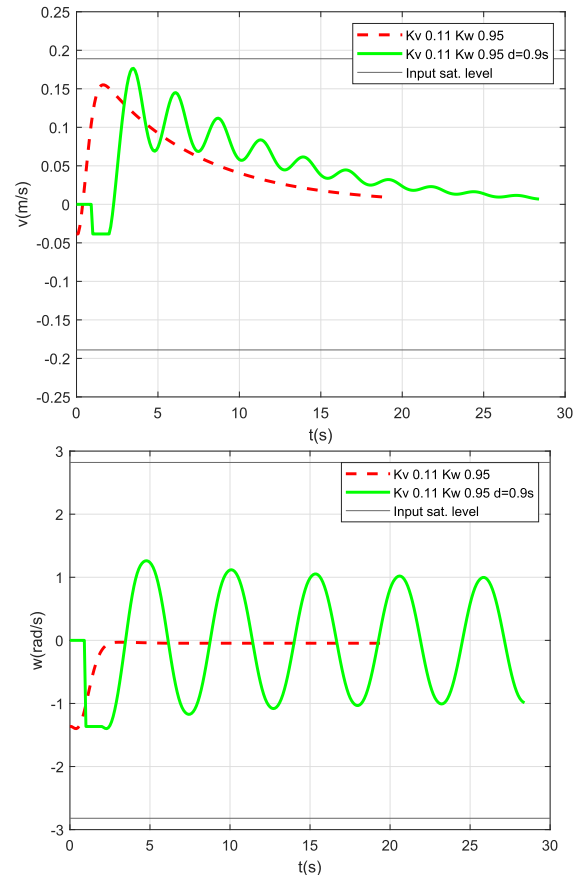


**FIGURE 4.** Simulation data: comparison of the time-evolution of the distance  $\rho(m)$  to the target position for different control gains  $K_V$ ,  $K_W$  with the designed values  $K_V = 0.11$  and  $K_W = 0.95$  in the presence of time delays.

to undesirable oscillations and even closed-loop instability caused by such delays. The control execution in all cases is forced to stop once the mobile robot is detected to be close to the target position. In Fig. 5, the time evolution of linear and angular velocities (control actions) are depicted for the designed control gains  $K_V = 0.11$  and  $K_W = 0.95$  considering the non-delayed and delayed case. It can be appreciated that the input saturation level is not reached during control execution, as expected from the two first constraints given in (8) (see Theorem 1).

### V. EXPERIMENTAL VALIDATION

Experimental tests were conducted on a physical system to validate the effectiveness of the proposed control synthesis method. The system was implemented using a LEGO platform and Arduino [34] (see Fig. 6). In our case, the mobile robot consists of a nonholonomic system such as depicted in Fig. 1. Two servo systems are implemented to obtain the desired angular speed at each wheel, namely  $w_L$  and  $w_R$  for



**FIGURE 5.** Simulation data: time-evolution of linear and angular velocities  $v$ ,  $w$  with the designed values  $K_V = 0.11$  and  $K_W = 0.95$ , together with the input saturation levels.



**FIGURE 6.** Two-wheeled mobile robot based on LEGO platform and Arduino.

left and right wheels respectively, where  $w_L < 13.5rad/s$  and  $w_R < 13.5rad/s$ . The angular velocity control systems for both wheels were adjusted to obtain a settling time around 0.4s, so their dynamics can be considered to be neglectable. It is easy to see that the angular velocities of left and right



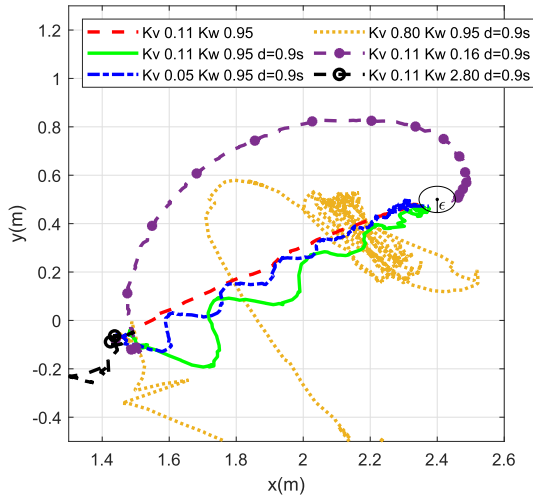


FIGURE 7. Experimental setup: system trajectories in the XY plane for different control gains including  $K_V = 0.11$  and  $K_W = 0.95$ .

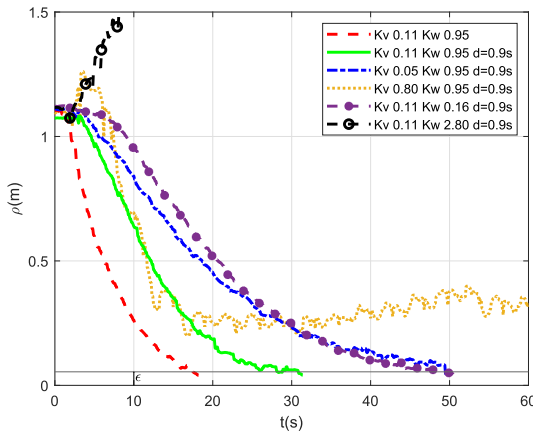


FIGURE 8. Experimental setup: time-evolution of the distance  $\rho(m)$  to the target position under different control gains including  $K_V = 0.11$  and  $K_W = 0.95$ .

wheels ( $w_L, w_R$ ) are related to  $v, w$  as:

$$\begin{bmatrix} w_L \\ w_R \end{bmatrix} = \begin{bmatrix} \frac{1}{r} & \frac{b}{r} \\ \frac{1}{r} & -\frac{b}{r} \end{bmatrix} \begin{bmatrix} v \\ w \end{bmatrix}, \quad (11)$$

where  $r$  is the wheel radius and  $b$  is the length of the axis between them (see Fig. 1). Therefore, neglecting the dynamics of the wheel servo systems, it is easy to see that  $w_L, w_R$  can directly be obtained by means of (11) from the proposed control laws  $w$  and  $v$  in (7). The parameter values for the prototype used in the experimental setup are  $r = 0.028m$  and  $b = 0.068m$  respectively, and the input saturation constraints are  $\bar{v} = 0.19m/s$  and  $\bar{w} = 2.82rad/s$ . The position  $(x, y)$  was measured using Marvelmind Indoor Positioning System [35] at 12 Hz.

The utilization of Arduino as an intermediary between LEGO and Marvelmind facilitates swift code development, as significant portions of the code designed for simulation can be seamlessly transferred to Arduino. However, some challenges arise when certain hardware components that are essential for the practical application are not inherently supported by Simulink. In such cases, additional implementa-

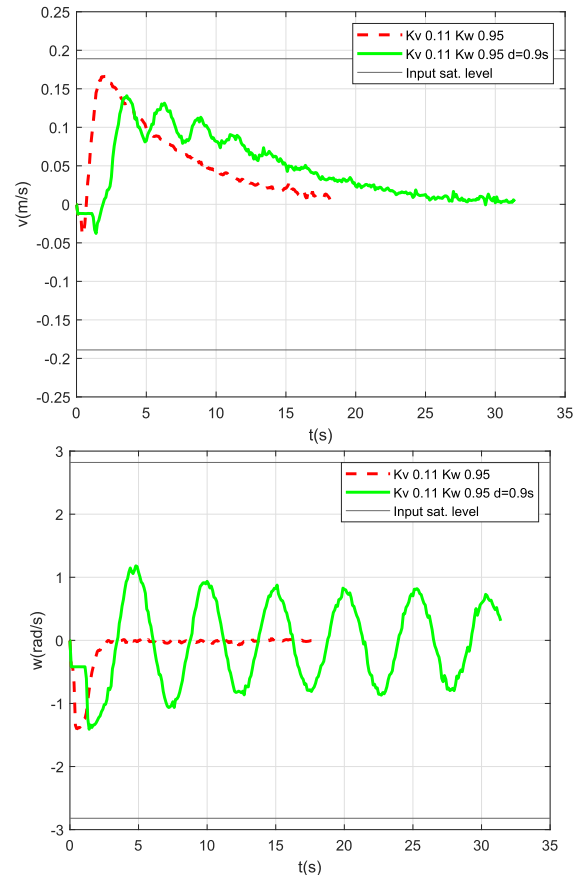


FIGURE 9. Experimental data: time-evolution of linear and angular velocities  $v, w$  with the designed values  $K_V = 0.11$  and  $K_W = 0.95$ , together with the input saturation levels.

tions become necessary to accommodate these requirements, such as integrating beacon readings or establishing communication with LEGO via USART.

For validation purposes, extra time-varying delays have been induced by software in order to increase the worst-case delay  $h_2$  until the settling time of  $30s$  is slightly overflowed with the control gains  $K_V = 0.11$  and  $K_W = 0.95$ , which were designed in Section IV. As a result, we obtain that the actual worst-case delay is  $h_2 = 0.9$ .

The meaning of Fig. 7 and Fig. 8 is respectively the same as Fig. 3 and Fig. 4, but with experimental results. From the figures, one can see that other choices for  $K_V, K_W$  give a slower response (if controller gains are smaller), overdamped response, or even instability (if controller gains are greater) in comparison to the designed  $K_V = 0.11$  and  $K_W = 0.95$ . This fact confirms the effectiveness of the control synthesis method. Additionally, the input saturation level is not reached during control execution, as can be deduced from the time evolution of  $v$  and  $w$  depicted in Fig. 9.

## VI. CONCLUSION

This work has presented a simple and low computational cost algorithm for the position control of nonholonomic mobile robots. The exponential convergence with guaranteed decay rate has theoretically been proved for any orientation

error. The key difficulties have been: (i) the construction of a suitable LKF candidate to reach a global exponential stability condition, in spite of the discontinuity of the kinematics model and the use of a smooth control law, (ii) the integration of some recent advances to reduce conservatism in performance analysis under time-varying delays, applied for the first time to nonholonomic systems, and (iii) the inclusion of extra LMI conditions to handle input saturation constraints. Finally, an experimental validation has been provided to show that the proposed method can effectively reduce the settling time to reach the target position in comparison with other choices for a certain worst-case delay.

**APPENDIX  
PROOF OF THEOREM 1**

First, let us define  $\rho_\alpha = \rho e^{\alpha t}$  in order to deal with exponential stabilization, where  $\alpha \geq 0$  is the decay rate. Then, given some scalars  $\gamma_1, \gamma_2, \gamma_3$  and symmetric matrices  $Q_1, Q_2 \in \mathcal{R}^2, Z_1, Z_2 \in \mathcal{R}^2 > 0$  and  $\bar{R} \in \mathcal{R}^6 > 0$ , the following nonquadratic Lyapunov-Krasovskii functional satisfying  $V > 0, \forall \rho_\alpha > 0, \phi, \theta \in \mathcal{R}$  is proposed:

$$\begin{aligned}
 V = & \frac{\gamma_1}{2} \rho_\alpha^2 + \frac{\gamma_2}{2} \rho_\alpha^2 \cos(\phi - \theta) + \frac{\gamma_3}{2} \rho_\alpha^2 \sin(\phi - \theta) + \xi_r^T \bar{R} \xi_r \\
 & + \int_{t-h_1}^t \xi^T(s) Q_1 \xi(s) ds \\
 & + h_1 \int_{-h_1}^0 \int_{t+s}^t \dot{\xi}^T(g) Z_1 \dot{\xi}(g) dg ds \\
 & + \int_{t-h_2}^{t-h_1} \xi^T(s) Q_2 \xi(s) ds \\
 & + \tau \int_{-h_2}^{-h_1} \int_{t+s}^t \dot{\xi}^T(g) Z_2 \dot{\xi}(g) dg ds
 \end{aligned} \tag{12}$$

where  $\tau = h_2 - h_1, \xi = [\xi_1^T \ \xi_2^T]^T$ , and

$$\begin{aligned}
 \xi_r = & \left[ \xi^T, \ \tilde{\eta}_1^T, \ \tilde{\eta}_2^T \right]^T, \\
 \tilde{\eta}_1(t) = & \int_{t-h_1}^t \xi(s) ds, \quad \tilde{\eta}_2(t) = \int_{t-h_2}^{t-h_1} \xi(s) ds
 \end{aligned} \tag{13}$$

with  $\xi_1, \xi_2$  defined as

$$\begin{aligned}
 \xi_1 = & \rho_\alpha \cos(\phi - \theta), \\
 \xi_2 = & \rho_\alpha \sin(\phi - \theta).
 \end{aligned} \tag{14}$$

The following conditions in  $V$  must be satisfied in order to ensure the global exponential convergence with decay rate  $\alpha$ : (i)  $V > 0$ , and (ii)  $\dot{V} < 0$  for all possible system trajectories. In addition, the conditions given in subsection VI-C must be satisfied in order not to reach the input saturation level:  $v \leq \bar{v}$  and  $w \leq \bar{w}$ . Hence, the proof is divided into the three corresponding subsections:

**A. PROOF OF  $V > 0$**

First, let us write  $V$  as

$$V = V^* + h_1 \int_{-h_1}^0 \int_{t+s}^t \dot{\xi}^T(g) Z_1 \dot{\xi}(g) dg ds$$

$$+ \tau \int_{-h_2}^{-h_1} \int_{t+s}^t \dot{\xi}^T(g) Z_2 \dot{\xi}(g) dg ds \tag{15}$$

where

$$\begin{aligned}
 V^* = & \sum_{i=1}^2 \sum_{j=1}^2 \mu_{1i} \mu_{2j} \left( \xi_r^T \hat{\Xi}_{ij} \xi_r \right) \\
 & + \int_{t-h_1}^t \xi^T(s) Q_1 \xi(s) ds + \int_{t-h_2}^{t-h_1} \xi^T(s) Q_2 \xi(s) ds
 \end{aligned} \tag{16}$$

and

$$\begin{aligned}
 \mu_{11} = & \frac{1}{2} (1 + \cos(\phi - \theta)), \quad \mu_{12} = 1 - \mu_{11}, \\
 \mu_{21} = & \frac{1}{2} (1 + \sin(\phi - \theta)), \quad \mu_{22} = 1 - \mu_{21}.
 \end{aligned} \tag{17}$$

with  $\hat{\Xi}_{ij}$  defined in (9). Now, applying Jensen's inequalities, one has that

$$\begin{aligned}
 & \int_{t-h_1}^t \xi^T(s) Q_1 \xi(s) ds \\
 & \geq \frac{1}{h_1} \left( \int_{t-h_1}^t \xi^T(s) \right) Q_1 \left( \int_{t-h_1}^t \xi(s) \right) \\
 & = \frac{1}{h_1} \xi_r^T (\text{diag}(0, Q_1, 0)) \xi_r, \\
 & \int_{t-h_2}^{t-h_1} \xi^T(s) Q_2 \xi(s) ds \\
 & \geq \frac{1}{h_2 - h_1} \left( \int_{t-h_2}^{t-h_1} \xi^T(s) \right) Q_2 \left( \int_{t-h_2}^{t-h_1} \xi(s) \right) \\
 & = \frac{1}{h_2 - h_1} \xi_r^T (\text{diag}(0, 0, Q_2)) \xi_r.
 \end{aligned} \tag{18}$$

Therefore, from (16) and (18), it can be deduced that

$$V^* > \sum_{i=1}^2 \sum_{j=1}^2 \beta_{1i} \beta_{2j} \left( \xi_r^T \Xi_{ij} \xi_r \right) \tag{19}$$

where  $\Xi_{ij}$  is given in (9). Taking into account that  $Z_1, Z_2 > 0$ , one can see from (15) that  $V^* > 0$  implies  $V > 0$ . Finally, applying convex sum properties considering that  $0 \leq \beta_{f1} \leq 1, 0 \leq \beta_{f2} \leq 1, f = 1, 2$  and  $\beta_{1f} + \beta_{2f} = 1$ , it can be deduced from (19) that  $V^* > 0$  if  $\Xi_{ij} > 0, i = 1, 2, j = 1, 2$ .

**B. PROOF OF  $V < 0$**

Time-derivative of  $V$  can be obtained as:

$$\begin{aligned}
 \dot{V} = & \gamma_1 \rho_\alpha \dot{\rho}_\alpha + \gamma_2 \rho_\alpha \cos(\phi - \theta) \dot{\rho}_\alpha \\
 & - \frac{\gamma_2}{2} \rho_\alpha^2 \sin(\phi - \theta) (\dot{\phi} - \dot{\theta}) \\
 & + \gamma_3 \rho_\alpha \sin(\phi - \theta) \dot{\rho}_\alpha \\
 & + \frac{\gamma_3}{2} \rho_\alpha^2 \cos(\phi - \theta) (\dot{\phi} - \dot{\theta}) \\
 & + \dot{\xi}_r^T \bar{R} \xi_r + \xi_r^T \bar{R} \dot{\xi}_r \\
 & + \xi^T(t-h_1) Q_1 \xi(t-h_1) - \xi^T Q_1 \xi \\
 & + \xi^T(t-h_2) Q_2 \xi(t-h_2) - \xi^T(t-h_1) Q_2 \xi(t-h_1) \\
 & + h_1^2 \dot{\xi}^T Z_1 \dot{\xi} - h_1 \int_{t-h_1}^t \dot{\xi}(s)^T Z_1 \dot{\xi}(s) ds
 \end{aligned}$$

$$+ \tau^2 \dot{\xi}^T Z_2 \dot{\xi} - \tau \int_{t-h_2}^{t-h_1} \dot{\xi}(s)^T Z_2 \dot{\xi}(s) ds \quad (20)$$

From the definition of  $\rho_\alpha = \rho e^{\alpha t}$ , the time-derivative renders

$$\dot{\rho}_\alpha = \dot{\rho} e^{\alpha t} + \alpha \rho_\alpha = \text{sat}(v^{(h)}) e^{\alpha t} \cos(\phi - \theta) + \alpha \rho_\alpha. \quad (21)$$

The fulfilment of the two first inequalities given in (8) implies that the input saturation level is never reached, that is to say,  $v < \bar{v}$  and  $w < \bar{w}$ , as proved later in Section VI-C. Hence, we have that:

$$\text{sat}(v^{(h)}) = v^{(h)}, \quad \text{sat}(w^{(h)}) = w^{(h)}. \quad (22)$$

Hence, replacing  $\dot{\rho}_\alpha$  from (21), together with  $\dot{\phi}, \dot{\theta}$  from (6) into the above expression, and taking into account (22), one has that:

$$\begin{aligned} \dot{V} = & \gamma_1 \rho_\alpha v^{(h)} e^{\alpha t} \cos(\phi - \theta) + \gamma_1 \alpha \rho_\alpha^2 \\ & + \gamma_2 \rho_\alpha v^{(h)} e^{\alpha t} \cos^2(\phi - \theta) + \gamma_2 \alpha \rho_\alpha^2 \cos(\phi - \theta) \\ & - \frac{\gamma_2}{2} \rho_\alpha^2 \sin(\phi - \theta) \left( -\frac{v^{(h)}}{\rho} \sin(\phi - \theta) - w^{(h)} \right) \\ & + \gamma_3 \rho_\alpha v^{(h)} e^{\alpha t} \cos(\phi - \theta) \sin(\phi - \theta) + \gamma_3 \alpha \rho_\alpha^2 \sin(\phi - \theta) \\ & + \frac{\gamma_3}{2} \rho_\alpha^2 \cos(\phi - \theta) \left( -\frac{v^{(h)}}{\rho} \sin(\phi - \theta) - w^{(h)} \right) \\ & + \dot{\xi}_r^T \bar{R} \dot{\xi}_r + \xi_r \bar{R} \dot{\xi}_r^T \\ & + \xi^T(t-h_1) Q_1 \xi(t-h_1) - \xi^T Q_1 \xi \\ & + \xi^T(t-h_2) Q_2 \xi(t-h_2) - \xi^T(t-h_1) Q_2 \xi(t-h_1) \\ & + h_1^2 \dot{\xi}^T Z_1 \dot{\xi} - h_1 \int_{t-h_1}^t \dot{\xi}(s)^T Z_1 \dot{\xi}(s) ds \\ & + \tau^2 \dot{\xi}^T Z_2 \dot{\xi} - \tau \int_{t-h_2}^{t-h_1} \dot{\xi}(s)^T Z_2 \dot{\xi}(s) ds. \end{aligned} \quad (23)$$

From (14) and (7) the following equivalences can be deduced:

$$\begin{aligned} \rho_\alpha^2 &= \xi_1^2 + \xi_2^2, \\ v^{(h)} e^{\alpha t} &= -K_v \xi_1^{(h)} e^{\alpha h(t)}. \end{aligned} \quad (24)$$

Taking into account (24) and  $\xi_1, \xi_2$  given in (14), the expression (23) can be rewritten as:

$$\begin{aligned} \dot{V} = & -K_v \gamma_1 e^{\alpha h(t)} \xi_1 \xi_1^{(h)} + \gamma_1 \alpha \left( \xi_1^2 + \xi_2^2 \right) \\ & - K_v \gamma_2 \xi_1 \xi_1^{(h)} e^{\alpha h(t)} \cos(\phi - \theta) \\ & + \gamma_2 \alpha \left( \xi_1^2 + \xi_2^2 \right) \cos(\phi - \theta) \\ & - \frac{K_v \gamma_2}{2} \xi_2 \xi_1^{(h)} e^{\alpha h(t)} \sin(\phi - \theta) - \frac{K_w \gamma_2}{2} \left( \frac{\rho_\alpha}{\rho^{(h)}} \right) \xi_2 \xi_2^{(h)} \\ & - K_v \gamma_3 \xi_1 \xi_1^{(h)} e^{\alpha h(t)} \sin(\phi - \theta) \\ & + \gamma_3 \alpha \left( \xi_1^2 + \xi_2^2 \right) \sin(\phi - \theta) \\ & - \frac{K_v \gamma_3}{2} \xi_2 \xi_1^{(h)} e^{\alpha h(t)} \cos(\phi - \theta) - \frac{K_w \gamma_3}{2} \left( \frac{\rho_\alpha}{\rho^{(h)}} \right) \xi_1 \xi_2^{(h)} \\ & + \dot{\xi}_r^T \bar{R} \dot{\xi}_r + \xi_r \bar{R} \dot{\xi}_r^T \\ & + \xi^T(t-h_1) Q_1 \xi(t-h_1) - \xi^T Q_1 \xi \end{aligned}$$

$$\begin{aligned} & + \xi^T(t-h_2) Q_2 \xi(t-h_2) - \xi^T(t-h_1) Q_2 \xi(t-h_1) \\ & + h_1^2 \dot{\xi}^T Z_1 \dot{\xi} - h_1 \int_{t-h_1}^t \dot{\xi}(s)^T Z_1 \dot{\xi}(s) ds \\ & + \tau^2 \dot{\xi}^T Z_2 \dot{\xi} - \tau \int_{t-h_2}^{t-h_1} \dot{\xi}(s)^T Z_2 \dot{\xi}(s) ds. \end{aligned} \quad (25)$$

The last integral term of the above expression can be decomposed as:

$$\begin{aligned} \tau \int_{t-h_2}^{t-h_1} \dot{\xi}(s)^T Z_2 \dot{\xi}(s) ds = & \tau \int_{t-h_1}^{t-h_1} \dot{\xi}(s)^T Z_2 \dot{\xi}(s) ds \\ & + \tau \int_{t-h_2}^{t-h(t)} \dot{\xi}(s)^T Z_2 \dot{\xi}(s) ds. \end{aligned} \quad (26)$$

Applying Wirtinger's inequality, the last two integral terms in (25) and (26) can respectively be bounded by:

$$\begin{aligned} -h_1 \int_{t-h_1}^t \dot{\xi}(s)^T Z_1 \dot{\xi}(s) ds & \leq -\chi_1^T \left( \mathcal{W}^T \bar{Z}_1 \mathcal{W} \right) \chi_1, \\ -\tau \int_{t-h(t)}^{t-h_1} \dot{\xi}(s)^T Z_2 \dot{\xi}(s) ds & \leq -\chi_{21}^T \left( \alpha_d(t) \mathcal{W}^T \bar{Z}_2 \mathcal{W} \right) \chi_{21}, \\ -\tau \int_{t-h_2}^{t-h(t)} \dot{\xi}(s)^T Z_2 \dot{\xi}(s) ds & \leq \\ -\chi_{22}^T \left( (1 - \alpha_d(t)) \mathcal{W}^T \bar{Z}_2 \mathcal{W} \right) & \chi_{22}, \end{aligned} \quad (27)$$

where  $\alpha_d(t) = \frac{h(t)-h_1}{\tau}$ , and

$$\begin{aligned} \chi_1 &= \left[ \xi^T(t) \quad \xi^T(t-h_1) \quad \frac{1}{h_1} \tilde{\eta}_1^T(t) \right]^T, \\ \chi_{21} &= \left[ \xi^T(t-h_1) \quad \xi^T(t-h(t)) \quad \frac{1}{h(t)-h_1} \tilde{\eta}_{21}^T(s) ds \right]^T, \\ \chi_{22} &= \left[ \xi^T(t-h(t)) \quad \xi^T(t-h_2) \quad \frac{1}{h_2-h(t)} \tilde{\eta}_{22}^T(s) ds \right]^T, \end{aligned} \quad (28)$$

where

$$\tilde{\eta}_{21}(s) = \int_{t-h(t)}^{t-h_1} \xi(s), \quad \tilde{\eta}_{22}(s) = \int_{t-h_2}^{t-h(t)} \xi(s). \quad (29)$$

Applying the Reciprocally Convex Lemma [23], we have that

$$\begin{aligned} & -\frac{1}{\alpha_d(t)} \chi_{21}^T \mathcal{W}^T \bar{Z}_2 \mathcal{W} \chi_{21} - \frac{1}{1 - \alpha_d(t)} \chi_{22}^T \mathcal{W}^T \bar{Z}_2 \mathcal{W} \chi_{22} \\ & \leq -\chi_2^T \bar{\mathcal{W}}^T \mathcal{T}(t) \bar{\mathcal{W}} \chi_2 \end{aligned} \quad (30)$$

where

$$\begin{aligned} \chi_2 &= \begin{bmatrix} \chi_{21} \\ \chi_{22} \end{bmatrix}, \quad \mathcal{T}(t) = \begin{bmatrix} \bar{Z}_2 + T_1(t) & T_2(t) \\ (*) & \bar{Z}_2 + T_3(t) \end{bmatrix}, \\ T_1(t) &= (1 - \alpha_d(t)) X_1, \quad T_3(t) = \alpha_d(t) X_2, \\ T_2(t) &= \alpha_d(t) Y_1 + (1 - \alpha_d(t)) Y_2, \end{aligned} \quad (31)$$

being  $X_1, X_2, Y_1, Y_2$  matrices satisfying  $\forall 0 \leq \alpha_d(t) \leq 1$ :

$$\begin{bmatrix} \bar{Z}_2 & 0 \\ 0 & \bar{Z}_2 \end{bmatrix} - \alpha_d(t) \begin{bmatrix} X_1 & Y_1 \\ (*) & 0 \end{bmatrix} - (1 - \alpha_d(t)) \begin{bmatrix} 0 & Y_2 \\ (*) & X_2 \end{bmatrix} \geq 0 \quad (32)$$



Therefore, we have from (26), (27) and (30) that

$$-\tau \int_{t-h_2}^{t-h_1} \dot{\xi}(s)^T Z_2 \dot{\xi}(s) \leq -\chi_2^T T(t) \chi_2 \quad (33)$$

Taking into account (27), (33) and (36), the time-derivative  $\dot{V}$  in (37) can be bounded as:

$$\begin{aligned} \dot{V} \leq & \xi^T \Pi_0(t) \xi + He \left( \xi^T \Pi_1(t) \xi^T (t-h(t)) \right) + He \left( \dot{\xi}_r^T \bar{R} \xi_r \right) \\ & + \xi^T Q_1 \xi - \xi^T (t-h_1) Q_1 \xi (t-h_1) \\ & + \xi^T Q_2 \xi - \xi^T (t-h_2) Q_2 \xi (t-h_2) \\ & + \dot{\xi}^T \tilde{Z} \dot{\xi} - \chi_1^T \left( \mathcal{W}^T \bar{Z}_1 \mathcal{W} \right) \chi_1 - \chi_2^T \bar{\mathcal{W}}^T T(t) \bar{\mathcal{W}} \chi_2 \end{aligned} \quad (34)$$

where

$$\Pi_0(t) = \begin{bmatrix} \Omega_0(t) & 0 \\ (*) & \Omega_0(t) \end{bmatrix}, \quad \Pi_1(t) = \begin{bmatrix} \Omega_1(t) & \Omega_2(t) \\ \Omega_3(t) & \Omega_4(t) \end{bmatrix}$$

and

$$\begin{aligned} \Omega_0(t) &= \alpha (\gamma_1 + \gamma_2 \cos(\phi - \theta) + \gamma_3 \sin(\phi - \theta)), \\ \Omega_1(t) &= -\frac{1}{2} K_v e^{\alpha h(t)} (\gamma_1 + \gamma_2 \cos(\phi - \theta) + \gamma_3 \sin(\phi - \theta)), \\ \Omega_2(t) &= -\frac{1}{4} K_w \gamma_3 e^{\alpha h(t)} \left( \frac{\rho_\alpha}{\rho_\alpha^{(h)}} \right), \\ \Omega_3(t) &= -\frac{1}{4} K_v (\gamma_2 \sin(\phi - \theta) + \gamma_3 \cos(\phi - \theta)) e^{\alpha h(t)}, \\ \Omega_4(t) &= -\frac{1}{4} K_w \gamma_2 e^{\alpha h(t)} \left( \frac{\rho_\alpha}{\rho_\alpha^{(h)}} \right). \end{aligned} \quad (35)$$

On the other hand, to deal with terms  $\dot{\xi}$  and  $\dot{\xi}_r$  in (34), note that time-derivative of (14) renders:

$$\begin{aligned} \dot{\xi}_1 &= sat(v^{(h)}) e^{\alpha t} + \alpha \rho_\alpha \cos(\phi - \theta) - \rho_\alpha \sin(\phi - \theta) sat(w^{(h)}), \\ \dot{\xi}_2 &= \alpha \rho_\alpha \sin(\phi - \theta) + \rho_\alpha \sin(\phi - \theta) sat(w^{(h)}). \end{aligned} \quad (36)$$

Applying (22), (14) and (7), the above expression can be rewritten as:

$$\begin{aligned} \dot{\xi}_1 &= -K_v \xi_1^{(h)} e^{\alpha h(t)} + \alpha \xi_1 + K_w \xi_2 \sin(\phi^{(h)} - \theta^{(h)}), \\ \dot{\xi}_2 &= \alpha \xi_2 - K_w \xi_2 \sin(\phi^{(h)} - \theta^{(h)}). \end{aligned} \quad (37)$$

The above expression in compact form renders:

$$\dot{\xi} = \Pi_{21}(t) \xi(t) + \Pi_{22}(t) \xi(t-h(t)) \quad (38)$$

where  $\Pi_{21}(t)$  and  $\Pi_{22}(t)$  are defined as:

$$\Pi_{21}(t) = \begin{bmatrix} \alpha & \Omega_5(t) \\ 0 & \Omega_6(t) \end{bmatrix}, \quad \Pi_{22}(t) = \begin{bmatrix} -K_v e^{\alpha h(t)} & 0 \\ 0 & 0 \end{bmatrix}$$

with

$$\begin{aligned} \Omega_5(t) &= K_w \sin(\phi^{(h)} - \theta^{(h)}), \\ \Omega_6(t) &= \alpha - K_w \sin(\phi^{(h)} - \theta^{(h)}) \end{aligned} \quad (39)$$

Now, let us define the augmented state vector:

$$\bar{\xi}^T(t) = [\xi^T(t), \quad \xi^T(t-h(t)), \quad \xi^T(t-h_1), \quad \xi^T(t-h_2),$$

$$\begin{aligned} & \frac{1}{h_1} \int_{t-h_1}^t \xi^T(s) ds, \quad \frac{1}{h(t)-h_1} \int_{t-h(t)}^{t-h_1} \xi^T(s) ds, \\ & \frac{1}{h_2-h(t)} \int_{t-h_2}^{t-h(t)} \xi^T(s) ds]^T \end{aligned} \quad (40)$$

Hence, we can write  $\xi$ ,  $\xi(t-h(t))$ ,  $\xi_r$ ,  $\dot{\xi}_r$ ,  $\chi_1$ ,  $\chi_2$  as a function of  $\bar{\xi} \equiv \bar{\xi}(t)$  as:

$$\begin{aligned} \xi &= E_1 \bar{\xi}, \quad \xi(t-h(t)) = E_2 \bar{\xi}, \\ \dot{\xi} &= \Pi_2(t) \bar{\xi}, \quad \xi_r = \Pi_3(t) \bar{\xi}, \quad \dot{\xi}_r = \Pi_4(t) \bar{\xi}, \\ \chi_1 &= \mathcal{D}_1 \bar{\xi}, \quad \chi_2 = \mathcal{D}_2 \bar{\xi} \end{aligned} \quad (41)$$

where

$$\begin{aligned} \Pi_2(t) &= \Pi_{21}(t) E_1 + \Pi_{22}(t) E_2, \\ \Pi_3(t) &= [E_1^T \quad h_1 E_5^T \quad \tau \alpha_d(t) E_6^T + \tau(1-\alpha_d(t)) E_7^T]^T, \\ \Pi_4(t) &= [\Pi_2^T(t) (E_1 - E_3)^T (E_3 - E_4)^T]^T. \end{aligned} \quad (42)$$

Therefore, we can write  $\dot{V}(t)$  as

$$\begin{aligned} \dot{V}(t) &= \bar{\xi}^T [E_1^T \Pi_0(t) E_1 + He (E_1^T \Pi_1(t) E_2) \\ & + He (\Pi_4^T(t) \bar{R} \Pi_3(t)) + E_1^T Q_1 E_1 - E_3^T Q_1 E_3 \\ & + E_1^T Q_2 E_1 - E_4^T Q_2 E_4 \\ & + \Pi_2^T(t) \tilde{Z} \Pi_2(t) - \mathcal{D}_1^T \mathcal{W}^T \bar{Z}_1 \mathcal{W} \mathcal{D}_1 \\ & - \mathcal{D}_2^T \bar{\mathcal{W}}^T T(t) \bar{\mathcal{W}} \mathcal{D}_2] \bar{\xi} \end{aligned} \quad (43)$$

The stability condition with exponential decay rate  $\alpha$  is satisfied if  $\dot{V} < 0$ , which is equivalent to

$$\begin{bmatrix} \Pi(t) & \Pi_2^T(t) \tilde{Z} \\ (*) & -\tilde{Z} \end{bmatrix} < 0 \quad (44)$$

where

$$\begin{aligned} \Pi(t) &= E_1^T \Pi_0(t) E_1 + He (E_1^T \Pi_1(t) E_2) \\ & + He (\Pi_4^T(t) \bar{R} \Pi_3(t)) + E_1^T Q_1 E_1 - E_3^T Q_1 E_3 \\ & + E_1^T Q_2 E_1 - E_4^T Q_2 E_4 \\ & - \mathcal{D}_1^T \mathcal{W}^T \bar{Z}_1 \mathcal{W} \mathcal{D}_1 - \mathcal{D}_2^T \bar{\mathcal{W}}^T T(t) \bar{\mathcal{W}} \mathcal{D}_2 \end{aligned} \quad (45)$$

Note that (44) contains the following nonlinear functions, that can be reformulated as:

$$\begin{aligned} \cos(\phi - \theta) &= \mu_{11}(t) \hat{\delta}_{11} + \mu_{12}(t) \hat{\delta}_{12}, \\ \sin(\phi - \theta) &= \mu_{21}(t) \hat{\delta}_{21} + \mu_{22}(t) \hat{\delta}_{22}, \\ \sin(\phi^{(h)} - \theta^{(h)}) &= \mu_{31}(t) \hat{\delta}_{31} + \mu_{32}(t) \hat{\delta}_{32}, \\ \left( \frac{\rho_\alpha}{\rho_\alpha^{(h)}} \right) &= \mu_{41}(t) \hat{\delta}_{41} + \mu_{42}(t) \hat{\delta}_{42}, \\ \alpha_d(t) &= \mu_{51}(t) \hat{\delta}_{51} + \mu_{52}(t) \hat{\delta}_{52}, \\ e^{\alpha h(t)} &= \mu_{51}(t) \hat{\delta}_{61} + \mu_{52}(t) \hat{\delta}_{62} \end{aligned} \quad (46)$$

where

$$\mu_{11} = \frac{1}{2} (1 - \cos(\phi - \theta)), \quad \mu_{12} = 1 - \mu_{11},$$

$$\begin{aligned} \mu_{21} &= \frac{1}{2} (1 - \sin(\phi - \theta)), & \mu_{22} &= 1 - \mu_{21}, \\ \mu_{31} &= \frac{1}{2} (1 - \sin(\phi^{(h)} - \theta^{(h)})), & \mu_{32} &= 1 - \mu_{31}, \\ \mu_{41} &= \frac{\rho_\alpha / \rho_\alpha^{(h)} - 1}{e^{-\alpha h_2} - 1}, & \mu_{42} &= 1 - \mu_{41}, \\ \mu_{51} &= \alpha_d(t), & \mu_{52} &= 1 - \mu_{51}, \\ \mu_{61} &= \frac{e^{\alpha h_2} - e^{\alpha h(t)}}{e^{\alpha h_2} - e^{\alpha h_1}}, & \mu_{62} &= 1 - \mu_{61}. \end{aligned} \quad (47)$$

Hence, we can write

$$\begin{aligned} \Omega_0(t) &= \alpha \sum_{i=1}^2 \sum_{j=1}^2 \mu_{1i} \mu_{2j} (\gamma_1 + \gamma_2 \hat{\delta}_{1i} + \gamma_3 \hat{\delta}_{2j}), \\ \Omega_1(t) &= -\frac{1}{2} K_v \sum_{i=1}^2 \sum_{j=1}^2 \sum_{p=1}^2 \mu_{1i} \mu_{2j} \mu_{6p} \\ &\quad \times (\gamma_1 + \gamma_2 \hat{\delta}_{1i} + \gamma_3 \hat{\delta}_{2j}) \hat{\delta}_{6p}, \\ \Omega_2(t) &= -\frac{1}{4} K_w \gamma_3 \sum_{m=1}^2 \sum_{p=1}^2 \mu_{4m} \mu_{6p} \hat{\delta}_{6p} \hat{\delta}_{4m}, \\ \Omega_3(t) &= -\frac{1}{4} K_v \sum_{i=1}^2 \sum_{j=1}^2 \sum_{p=1}^2 \mu_{1i} \mu_{2j} \mu_{6p} (\gamma_2 \hat{\delta}_{2j} + \gamma_3 \hat{\delta}_{1i}) \hat{\delta}_{6p}, \\ \Omega_4(t) &= -\frac{1}{4} K_w \gamma_2 \sum_{m=1}^2 \sum_{p=1}^2 \mu_{4m} \mu_{6p} \hat{\delta}_{6p} \hat{\delta}_{4m}, \\ \Omega_5(t) &= K_w \sum_{k=1}^2 \mu_{3k} \hat{\delta}_{k3}, & \Omega_6(t) &= \alpha - K_w \sum_{k=1}^2 \mu_{3k} \hat{\delta}_{k3}, \end{aligned} \quad (48)$$

and therefore

$$\begin{aligned} \Pi_0(t) &= \sum_{i=1}^2 \sum_{j=1}^2 \mu_{1i} \mu_{2j} \hat{\Pi}_{0,ij}, \\ \Pi_1(t) &= \sum_{i=1}^2 \sum_{j=1}^2 \sum_{m=1}^2 \sum_{p=1}^2 \mu_{1i} \mu_{2j} \mu_{4m} \mu_{6p} \hat{\Pi}_{1,ijmp}, \\ \Pi_2(t) &= \sum_{k=1}^2 \sum_{p=1}^2 \mu_{3k} \mu_{6p} \hat{\Pi}_{2,kp}, & \Pi_3(t) &= \sum_{n=1}^2 \mu_{5n} \hat{\Pi}_{3,n}, \\ \Pi_4(t) &= \sum_{k=1}^2 \sum_{p=1}^2 \mu_{3k} \mu_{6p} \hat{\Pi}_{4,kp}, & \mathcal{T}(t) &= \sum_{n=1}^2 \mu_{5n} \hat{T}_n. \end{aligned} \quad (49)$$

The above inequality can equivalently be expressed as:

$$\begin{aligned} &\sum_{i=1}^2 \sum_{j=1}^2 \sum_{k=1}^2 \sum_{m=1}^2 \sum_{n=1}^2 \sum_{p=1}^2 \mu_{1i} \mu_{2j} \mu_{3k} \mu_{4m} \mu_{5n} \mu_{6p} \\ &\quad \times \begin{bmatrix} \hat{\Pi}_{ijkmp} & \hat{\Pi}_{2,kp}^T \tilde{Z} \\ (*) & -\tilde{Z} \end{bmatrix} < 0, \end{aligned} \quad (50)$$

Finally, noting that the auxiliary functions  $\mu_{f1} = \mu_{f1}(t)$ ,  $\mu_{f2} = \mu_{f2}(t)$ ,  $f = 1, \dots, 6$  given in (47) satisfy  $0 \leq \mu_{f1} \leq$

$1$ ,  $0 \leq \mu_{f2} \leq 1$  and  $\mu_{f1} + \mu_{f2} = 1, \forall t \geq 0$ . Hence, by the convex sum properties of (50), it can be seen that the fulfilment of the matrix inequalities given in the fourth line of (8) implies that (50) is true, leading to  $\lim_{t \rightarrow \infty} \xi = 0$ . By the definition of  $\xi$  and  $\xi_1, \xi_2$  in (14), this last condition leads to  $\lim_{t \rightarrow \infty} \rho_\alpha = 0$  for any orientation error  $\phi - \theta$ , which guarantees the global exponential convergence with decay rate  $\alpha$  in the sense of Definition 1, recalling that  $\rho_\alpha = \rho e^{\alpha t}$ .

### C. PROOF OF NO INPUT SATURATION

First, it is trivial to see from (7) that  $K_w < \bar{w}$  ensures that  $w < \bar{w}$ . To guarantee that  $v$  satisfies  $v < \bar{v}, \forall t \geq 0$ , we assume that the initial distance from the mobile robot to the target position is time-constant and known:  $\rho(t \leq 0) = \rho_0$ , where  $\rho_0$  is the initial distance. The initial orientation error, denoted as  $\phi_0 - \theta_0$ , is also assumed to be time-constant but not necessarily known. Hence, two conditions must be satisfied:

- The initial state vector  $\xi_0 = [\xi_{10}, \xi_{20}]^T$  at time instant  $t \leq 0$  with

$$\xi_{10} = \rho_0 \cos(\phi_0 - \theta_0), \quad \xi_{20} = \rho_0 \sin(\phi_0 - \theta_0) \quad (51)$$

must be contained into a circle of radius  $\bar{\rho}_0$ , which is true if

$$V_0 < \left( \frac{1}{\bar{\rho}_0^2} \right) \xi_0^T \xi_0, \quad V_0 \equiv V|_{\rho=\rho_0, \phi=\phi_0, \theta=\theta_0, t=0} \quad (52)$$

taking into account that  $\xi_{10}^2 + \xi_{20}^2 = \rho_0^2$ . From (12) we have

$$\begin{aligned} V_0 &= \frac{\gamma_1}{2} \rho_0^2 + \frac{\gamma_2}{2} \rho_0^2 \cos(\phi_0 - \theta_0) \\ &\quad + \frac{\gamma_3}{2} \rho_0^2 \sin(\phi_0 - \theta_0) + \xi_{r0}^T \bar{R} \xi_{r0} \\ &\quad + \int_{t-h_1}^t \xi_0^T(s) Q_1 \xi_0(s) ds \\ &\quad + h_1 \int_{-h_1}^0 \int_{t+s}^t \xi_0^T(g) Z_1 \dot{\xi}_0(g) dg ds \end{aligned} \quad (53)$$

$$\begin{aligned} &+ \int_{t-h_2}^{t-h_1} \xi_0^T(s) Q_2 \xi_0(s) ds \\ &+ \tau \int_{-h_2}^{t-h_1} \int_{t+s}^t \xi_0^T(g) Z_2 \dot{\xi}_0(g) dg ds \end{aligned} \quad (54)$$

Recalling that  $\xi_0$  is time-constant  $\forall t \leq 0$ , then  $\dot{\xi}_0(g) = 0, -h_2 \leq g \leq 0$ , and the above expression is equivalent to

$$\begin{aligned} V_0 &= \frac{\gamma_1}{2} \rho_0^2 + \frac{\gamma_2}{2} \rho_0^2 \cos(\phi_0 - \theta_0) \\ &\quad + \frac{\gamma_3}{2} \rho_0^2 \sin(\phi_0 - \theta_0) + \xi_{r0}^T \bar{R} \xi_{r0} \\ &\quad + \int_{t-h_1}^t \xi_0^T(s) Q_1 \xi_0(s) ds + \int_{t-h_2}^{t-h_1} \xi_0^T(s) Q_2 \xi_0(s) ds, \end{aligned} \quad (55)$$

where  $\xi_{r0} = \left[ \xi_0, \int_{t-h_1}^t \xi_0(s)ds, \int_{t-h_2}^{t-h_1} \xi_0(s)ds \right]^T$ . Recalling again that  $\xi_0(s) = 0, -h_2 \leq s \leq 0$  and considering the above expression, the inequality (52) renders:

$$V_0 = \xi_0^T \left( \sum_{i=1}^2 \sum_{j=1}^2 \beta_{1i}\beta_{2j} \hat{\Xi}_{ij} + R_1 + h_1^2 R_4 + \tau^2 R_6 + h_1 (R_2 + R_2^T) + \tau (R_3 + R_3^T) + h_1 \tau (R_5 + R_5^T) + h_1 Q_1 + \tau Q_2 \right) \xi_0 < \left( \frac{1}{\bar{\rho}_0^2} \right) \xi_0^T \xi_0, \quad (56)$$

where

$$\beta_{11} = \frac{1}{2} (1 + \cos(\phi_0 - \theta_0)), \quad \beta_{12} = 1 - \beta_{11},$$

$$\beta_{21} = \frac{1}{2} (1 + \sin(\phi_0 - \theta_0)), \quad \beta_{22} = 1 - \beta_{21}.$$

The above inequality (56) holds if

$$\sum_{i=1}^2 \sum_{j=1}^2 \beta_{1i}\beta_{2j} (\hat{\Xi}_{ij} + R_1 + h_1^2 R_4 + \tau^2 R_6 + h_1 (R_2 + R_2^T) + \tau (R_3 + R_3^T) + h_1 \tau (R_5 + R_5^T) + h_1 Q_1 + \tau Q_2) < \frac{1}{\bar{\rho}_0^2}. \quad (57)$$

- The velocity must not reach  $\bar{v}$  for all initial distances to the target position satisfying  $\rho_0 \leq \bar{\rho}_0$  regardless of the initial orientation error  $\phi_0 - \theta_0$ . Provided that the condition given above is satisfied, we have to ensure that  $v \leq \bar{v}$  for all system trajectories contained in the level set  $V < 1$ . Hence, the following condition is established:

$$(1 - V) + \lambda_d (v^T v - \bar{v}^2) < 0 \quad (58)$$

for any scalar  $\lambda_d$ . Choosing  $\lambda_d = 1/\bar{v}^2$ , we have that

$$-V + \frac{1}{\bar{v}^2} v^T v < 0 \quad (59)$$

Taking into account that  $Z_1 > 0, Z_2 > 0$ , from the structure of  $V$  in (12) it is true that

$$-V < -\xi_r^T \left( \sum_{i=1}^2 \sum_{j=1}^2 \mu_{1i}\mu_{2j} \Xi_{ij} \right) \xi_r. \quad (60)$$

Therefore, applying (60) and noting that  $v^T v = \xi_1^T K_v^T K_v \xi_1 e^{-2\alpha t} < \xi_1^T K_v^T K_v \xi_1$ , we have that

$$-V + \frac{1}{\bar{v}^2} v^T v < -\xi_r^T \left( \sum_{i=1}^2 \sum_{j=1}^2 \mu_{1i}\mu_{2j} \Xi_{ij} \right) \xi_r + \frac{1}{\bar{v}^2} \xi_r^T \bar{K}_v^T \bar{K}_v \xi_r < 0. \quad (61)$$

The inequality (61) is true  $\forall \xi_r$  if

$$\sum_{i=1}^2 \sum_{j=1}^2 \mu_{1i}\mu_{2j} \left( -\Xi_{ij} + \frac{1}{\bar{v}^2} \bar{K}_v^T \bar{K}_v \right) < 0. \quad (62)$$

Applying Schur Complement, the above inequality is true if

$$\sum_{i=1}^2 \sum_{j=1}^2 \mu_{1i}\mu_{2j} \begin{bmatrix} \Xi_{ij} & \bar{K}_v^T \\ \bar{K}_v & \bar{v}^2 \end{bmatrix} > 0. \quad (63)$$

The fulfilment of both conditions (57) and (63) implies that if the initial distance to the target position is less or equal than  $\rho_0$ , then the velocity  $v$  does not exceed  $\bar{v}$  for any  $t \geq 0$ . Finally, it is easy to see that the inequalities (57) and (63) hold if the two first inequalities depicted in (8) are true.

### DISCLOSURE STATEMENT

No potential conflict of interest was reported by the author(s).

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