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# Compensation of state-dependent state delay for nonlinear systems

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### ABSTRACT

We extend the technique for compensating state-dependent delays from systems with delayed inputs to systems with delayed states. We focus on predictor-feedback design for nonlinear systems in the strictfeedback form, having a state-dependent state delay on the virtual input. The two key challenges are the definition of the predictor state and the fact that the predictor design does not follow immediately from the delay-free design. We resolve these challenges and we establish asymptotic stability of the resulting infinite-dimensional nonlinear system for general nonnegative-valued delay functions of the state. Due to an inherent limitation on the delay rate, and since the delay rate depends on the state, we obtain only regional stability results. However, for forward-complete systems, we provide an estimate of the region of attraction in the state space of the infinite-dimensional system. We finally provide two examples, including an example of stabilization of a cooling system.

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#### 1. Introduction

State-dependent state delays appear in many engineering applications. Examples include milling processes [1], engine cooling systems [2], irrigation channels [3], network congestion control [4], population dynamics [5], supply networks [6,7] and automatic landing systems [8].

Predictor-based techniques are an indispensable part of the control design toolbox [9], for unstable linear plants with constant delays affecting the input [10-13] or simultaneously affecting inputs and states [14-17]. Various control schemes also exist for nonlinear systems with constant delays affecting the input [18-20] or state [21-23]. Yet, extensions of the predictor-feedback design to nonlinear systems with constant input delay had not been developed until recently [24,25]. Although in [26] (see also [27]), a predictor-based controller for unstable linear plants with time-varying input delay is developed, only recently a Lyapunov function was provided [28]. Finally, although nonlinear systems with simultaneous time-varying input and state delays are considered in [29], predictor-like designs for nonlinear systems with time-varying input delays [30] or simultaneous input and state delays [31] were developed recently. In [32], we introduced a technique for compensating state-dependent delays on the input of a nonlinear system. In this paper, we generalize this technique to systems that include state-dependent delays on the states of the system.

We consider forward-complete systems that are globally stabilizable in the absence of the delay. We then "backstep" one state-dependent integrator and design a predictor-based control law for the overall system, using prediction intervals that depend on the current value of the state (Section 2). Using an invertible infinite-dimensional backstepping transformation we derive explicit bounds for the norm of the closed-loop system. Due to the fundamental limitation of the allowable magnitude of the delay function's gradient (the control signal never reaches the plant if the delay rate is larger than one) we use these bounds to estimate the region of attraction of the proposed controller (Section 3). Two simulation examples illustrate the application of the control design (Sections 4 and 5).

Notation: we use the common definition of class  $\mathscr{K}$ ,  $\mathscr{K}_{\infty}$  and  $\mathscr{KL}$  functions from [33]. For an *n*-vector, the norm  $|\cdot|$  denotes the usual Euclidean norm. We say that a function  $\xi : \mathbb{R}_+ \times (0, 1) \mapsto \mathbb{R}_+$ belongs to class  $\mathscr{K}\mathscr{C}$  if it is of class  $\mathscr{K}$  with respect to its first argument for each value of its second argument and continuous with respect to its second argument. It belongs to class  $\mathscr{KC}_{\infty}$  if it is in  $\mathscr{K}\mathscr{C}$  and also in  $\mathscr{K}_{\infty}$  with respect to its first argument.

## 2. Problem formulation and controller design

We consider the following system

$$X_1(t) = f_1(t, X_1(t), X_2(t - D(X_1(t))))$$
(1)

$$X_2(t) = f_2(t, X_1(t), X_2(t)) + U(t),$$
(2)

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where  $X_1 \in \mathbb{R}^n$ ,  $X_2$ ,  $U \in \mathbb{R}$  and  $t \ge t_0 \ge 0$ . We assume that  $f_1 : [t_0, \infty) \times \mathbb{R}^{n+1} \mapsto \mathbb{R}^n$  is locally Lipschitz with  $f_1(t, 0, 0) = 0$  for all  $t > t_0$  and that there exists a class  $\mathscr{K}_{\infty}$  function  $\hat{\alpha}$  such that

$$|f_1(t, X_1, X_2)| \le \hat{\alpha} \left( |X_1| + |X_2| \right), \quad \text{for all } t \ge t_0.$$
(3)

We further assume that  $f_2 : [t_0, \infty) \times \mathbb{R}^{n+1} \mapsto \mathbb{R}$  is locally Lipschitz with respect to  $(X_1, X_2) \in \mathbb{R}^{n+1}$  with  $f_2(t, 0, 0) = 0$  for all  $t \ge t_0$ . The goal of the paper is to show that for (1), (2) there exist functions  $P(\theta)$  and  $\sigma(\theta)$ , where  $t - D(X_1(t)) \le \theta \le t$ , such that the controller

$$U(t) = -f_{2}(t, X_{1}(t), X_{2}(t)) - c_{2}(X_{2}(t) - \kappa(\sigma(t), P_{1}(t))) + \frac{\frac{\partial \kappa(\sigma, P_{1})}{\partial \sigma} + \frac{\partial \kappa(\sigma, P_{1})}{\partial P_{1}}f_{1}(\sigma(t), P_{1}(t), X_{2}(t))}{1 - \nabla D(P_{1}(t))f_{1}(\sigma(t), P_{1}(t), X_{2}(t))},$$
(4)

where  $c_2$  is an arbitrary positive constant and

$$P_{1}(\theta) = X_{1}(t) + \int_{t-D(X_{1}(t))}^{\theta} \frac{f_{1}(\sigma(s), P_{1}(s), X_{2}(s)) ds}{1 - \nabla D(P_{1}(s)) f_{1}(\sigma(s), P_{1}(s), X_{2}(s))}, \\ t - D(X_{1}(t)) \le \theta \le t$$
(5)

$$\sigma(\theta) = \theta + D(P_1(\theta)), \quad t - D(X_1(t)) \le \theta \le t,$$
(6)

for  $t \ge t_0$ , compensates the state-dependent state delay and achieves asymptotic stability of the resulting closed-loop system. We refer to the quantity  $P_1(\theta)$  given in (5) as "predictor" since  $P_1(t)$ is the  $D(P_1(t))$  time units ahead predictor of  $X_1(t)$ , i.e.,  $P_1(t) = X_1(t + D(P_1(t)))$ . This fact can be seen as follows. Differentiating relation (5) with respect to  $\theta$  and setting  $\theta = t$  we get

$$\frac{dP_1(t)}{dt} = \frac{f_1(\sigma(t), P_1(t), X_2(t))}{1 - \nabla D(P_1(t)) f_1(\sigma(t), P_1(t), X_2(t))}.$$
(7)

Performing a change of variables  $\tau = \sigma(t)$  in the ODE for  $X_1(\tau)$  given by  $\frac{dX_1(\tau)}{d\tau} = f_1(\tau, X_1(\tau), X_2(\tau - D(X_1(\tau))))$ , we have that

$$\frac{dX_1(\sigma(t))}{dt} = \frac{d\sigma(t)}{dt} f_1(\sigma(t), X_1(\sigma(t)), X_2(t)).$$
(8)

From (8) one observes that  $P_1(t)$  satisfies the same ODE in t as  $X_1(\sigma(t))$  because from (6) to (8) it follows that

$$\frac{d\sigma(\theta)}{d\theta} = \frac{1}{1 - \nabla D(X_1(\sigma(\theta)))f_1(\sigma(\theta), X_1(\sigma(\theta)), X_2(\theta))},$$
  
$$t - D(X_1(t)) \le \theta \le t,$$
(9)

provided that  $P_1(t) = X_1(\sigma(t))$ . Since from (5) for  $t = t_0$  and  $\theta = t_0 - D(X_1(t_0))$  it follows that  $P_1(t_0 - D(X_1(t_0))) = X_1(t_0)$ , by defining

$$\phi(t) = t - D(X_1(t)), \quad t \ge t_0, \tag{10}$$

$$\sigma(\theta) = \phi^{-1}(\theta), \quad t - D(X_1(t)) \le \theta \le t, \tag{11}$$

we have that  $P_1(t_0) = X_1(\sigma(t_0))$ . Noting from (10) and (11) that  $D(X_1(\sigma(t))) = \sigma(t) - t$ , differentiating this relation, we get (9). Comparing (7) with (8) we conclude with the help of (9) that indeed  $P_1(t) = X_1(\sigma(t))$  for all  $t \ge t_0$ .

Motivated by the need to keep the denominator in (5) and (9) positive, throughout the paper we consider the condition on the solutions which is given by

$$\mathscr{G}_{c}: \nabla D\left(P_{1}(\theta)\right) f_{1}\left(\sigma\left(\theta\right), P_{1}(\theta), X_{2}(\theta)\right) < c,$$
  
for all  $\theta \ge t_{0} - D\left(X_{1}(t_{0})\right),$  (12)

for  $c \in (0, 1]$ . We refer to  $\mathscr{G}_1$  as the *feasibility condition* of the controller (4)–(5).

## 3. Stability analysis for forward-complete systems

Throughout the section, we make the following assumptions concerning the plant (1)-(2):

**Assumption 1.**  $D \in C^1(\mathbb{R}^n; \mathbb{R}_+)$  and  $\nabla D$  is locally Lipschitz.<sup>1</sup>

**Assumption 2.** There exist a smooth positive definite function *R* and class  $\mathscr{K}_{\infty}$  functions  $\alpha_1$ ,  $\alpha_2$  and  $\alpha_3$  such that for the plant  $\dot{X} = f_1(t, X, \omega)$ , the following hold

$$\alpha_1(|X|) \le R(t, X) \le \alpha_2(|X|) \tag{13}$$

$$\frac{\partial R(t,X)}{\partial t} + \frac{\partial R(t,X)}{\partial X} f_1(t,X,\omega) \le R(t,X) + \alpha_3(|\omega|), \qquad (14)$$

for all  $X, \omega \in \mathbb{R}^{n+1}$  and  $t \ge t_0$ .

Assumption 2 guarantees that the plant  $\dot{X} = f_1(t, X, \omega)$  with  $\omega$  as input is forward-complete.

**Assumption 3.** There exist functions  $\kappa \in C^1([t_0, \infty) \times \mathbb{R}^n; \mathbb{R})$ and  $\hat{\rho} \in \mathscr{K}_{\infty}$ , such that the plant  $\dot{X}(t) = f_1(t, X(t), \kappa(t, X(t)) + \omega(t))$  is input-to-state stable with respect to  $\omega$  and  $\kappa$  is uniformly bounded with respect to its first argument, that is,

$$|\kappa(t,X)| \le \hat{\rho}(|X|), \quad \text{for all } t \ge t_0.$$
(15)

**Theorem 1.** Consider the plant (1)-(2) together with the control law (4)-(6). Under Assumptions 1–3, there exist a class  $\mathscr{K}$  function  $\xi_{\text{RoA}}$ , a class  $\mathscr{KL}$  function  $\beta$  and a class  $\mathscr{KC}_{\infty}$  function  $\xi_1$  such that for all initial conditions for which  $X_2$  is locally Lipschitz on the interval  $[t_0 - D(X_1(t_0)), t_0]$  and which satisfy

$$|X_{1}(t_{0})| + \sup_{t_{0} - D(X_{1}(t_{0})) \le \theta \le t_{0}} |X_{2}(\theta)| < \xi_{\text{RoA}}(c),$$
(16)

for some 0 < c < 1, there exists a unique solution to the closed-loop system with  $X_1 \in C^1[t_0, \infty), X_2 \in C^1(t_0, \infty)$ , and

$$\begin{aligned} |X_{1}(t)| &+ \sup_{t - D(X_{1}(t)) \le \theta \le t} |X_{2}(\theta)| \\ &\le \beta \left( \xi_{1} \left( |X_{1}(t_{0})| + \sup_{t_{0} - D(X_{1}(t_{0})) \le \theta \le t_{0}} |X_{2}(\theta)|, c \right), t - t_{0} \right), (17) \end{aligned}$$

for all  $t \ge t_0$ . Furthermore, there exists a class  $\mathscr{K}$  function  $\delta^*$ , such that for all  $t \ge t_0$  the following holds

$$D(X_1(t)) \le D(0) + \delta^*(c)$$
(18)

$$\left| \hat{D}\left( X_{1}(t) \right) \right| \leq c. \tag{19}$$

The proof of Theorem 1 is based on Lemmas 1–8 which are presented next.

**Lemma 1** (Backstepping Transform of the Delayed State). The infinite-dimensional backstepping transformation of the state  $X_2$  defined by

$$Z_{2}(\theta) = X_{2}(\theta) - \kappa \left(\sigma(\theta), P_{1}(\theta)\right),$$
  
$$t - D\left(X_{1}(t)\right) \le \theta \le t,$$
(20)

together with the predictor-based control law given in relations (4)-(5) transform system (1)-(2) to the "target system" given by

$$\dot{X}_1(t) = f_1(t, X_1(t), \kappa(t, X_1(t)) + Z_2(t - D(X_1(t))))$$
(21)

$$\dot{Z}_2(t) = -c_2 Z_2(t).$$
 (22)

<sup>&</sup>lt;sup>1</sup> To ensure uniqueness of solutions.



**Fig. 1.** Interconnections between the predictor states  $P_1$  and  $\Pi_1$  with the transformations  $Z_2$  and  $X_2$  in (20) and (23). The direct backstepping transformation is defined as  $(X_1(t), X_2(\theta)) \mapsto (X_1(t), Z_2(\theta))$  and is given in (20), where  $P_1(\theta)$  is given as a function of  $X_1(t)$  and  $X_2(\theta)$  through relation (5). Analogously, the inverse transformation is defined as  $(X_1(t), Z_2(\theta)) \mapsto (X_1(t), X_2(\theta))$  and is given in (23) where  $\Pi_1(\theta)$  is given as a function of  $X_1(t)$  and  $Z_2(\theta)$  through relation (23).

**Proof.** Using (1) and the facts that  $P_1(t - D(X_1(t))) = X_1(t)$  and  $\sigma(t - D(X_1(t))) = t$ , which are immediate consequences of (5) and (6), we get (21). Setting  $\theta = t$  in (20) and taking the derivative with respect to *t* of the resulting equation we get (22) using (4), (7) and (9).  $\Box$ 

**Lemma 2** (Inverse Backstepping Transform). The inverse of the infinite-dimensional backstepping transformation defined in (20) is given by

$$X_{2}(\theta) = Z_{2}(\theta) + \kappa \left(\sigma(\theta), \Pi_{1}(\theta)\right),$$
  
$$t - D\left(X_{1}(t)\right) \le \theta \le t,$$
(23)

where  $\Pi_1(\theta)$  is defined in Box I.

**Proof.** By direct verification, noting also that  $\Pi_1(\theta) = P_1(\theta)$  for all  $t - D(X_1(t)) \le \theta \le t$ , where  $\Pi_1(\theta)$  is driven by the transformed state  $Z_2(\theta)$ , whereas  $P_1(\theta)$  is driven by the state  $X_2(\theta)$  for  $t - D(X_1(t)) \le \theta \le t$ . See Fig. 1.  $\Box$ 

**Lemma 3** (Stability Estimate for Target System). There exists a class  $\mathscr{KL}$  function  $\beta^*$  such that for all solutions of system (1), (2) satisfying (12) for 0 < c < 1, the following holds for all  $t \ge t_0$ 

$$|X_{1}(t)| + \sup_{t-D(X_{1}(t)) \le \theta \le t} |Z_{2}(\theta)|$$
  
$$\leq \beta^{*} \left( |X_{1}(t_{0})| + \sup_{t_{0}-D(X_{1}(t_{0})) \le \theta \le t_{0}} |Z_{2}(\theta)|, t-t_{0} \right).$$
(25)

**Proof.** Solving (22), and using the facts that  $\sigma(t_0) = t_0 + D(X_1 (\sigma(t_0)))$  and that  $\phi(t)$  is increasing for all  $t \ge t_0$  we get

$$\sup_{t-D(X_1(t)) \le \theta \le t} |Z_2(\theta)|$$

$$\le \sup_{t_0 - D(X_1(t_0)) \le \theta \le t_0} |Z_2(\theta)|$$

$$\times e^{-c_2(t-D(X_1(t)) - t_0)}, \text{ for all } t \ge \sigma(t_0),$$
(26)

where we have also used the trivial inequality  $|Z_2(t_0)| \le \sup_{t_0-D(X_1(t_0)) \le \theta \le t_0} |Z_2(\theta)|$ . Similarly, for all  $t_0 \le t \le \sigma(t_0)$  we get

$$\sup_{t-D(X_1(t)) \le \theta \le t} |Z_2(\theta)| \le \sup_{t_0 - D(X_1(t_0)) \le \theta \le t_0} |Z_2(\theta)| + \sup_{t_0 \le \theta \le t \le \sigma(t_0)} |Z_2(\theta)|,$$
(27)

#### and hence, combining (27) with (22), we get

$$\sup_{t-D(X_1(t)) \le \theta \le t} |Z_2(\theta)| \le 2 \sup_{t_0 - D(X_1(t_0)) \le \theta \le t_0} |Z_2(\theta)|,$$
  
for all  $t_0 \le t \le \sigma(t_0)$ . (28)

Therefore, using (26) and (28) we get

$$\sup_{t-D(X_{1}(t)) \le \theta \le t} |Z_{2}(\theta)|$$
  

$$\le 2 \sup_{t_{0}-D(X_{1}(t_{0})) \le \theta \le t_{0}} |Z_{2}(\theta)| e^{c_{2}(D(0)+\delta_{1}(|X_{1}(t)|))} e^{-c_{2}(t-t_{0})},$$
  
for all  $t \ge t_{0}.$ 
(29)

Under Assumption 3 and [34], (see also [35,36]), there exist class  $\mathscr{KL}$  function  $\hat{\beta}$  and class  $\mathscr{K}$  function  $\hat{\gamma}$  such that

$$\begin{aligned} |X_1(t)| &\leq \hat{\beta} \left( |X_1(s)|, t-s \right) \\ &+ \hat{\gamma} \left( \sup_{s \leq \tau \leq t} |Z_2\left(\tau - D\left(X_1(\tau)\right)\right)| \right), \\ &\text{for all } t \geq s \geq t_0. \end{aligned}$$
(30)

Setting  $s = t_0$  we have that

$$\begin{aligned} |X_1(t)| &\leq \hat{\beta} \left( |X_1(t_0)|, 0 \right) \\ &+ \hat{\gamma} \left( \sup_{\phi(t_0) \leq \theta \leq t_0} |Z_2(\theta)| + \sup_{t_0 \leq \theta \leq \phi(t)} |Z_2(\theta)| \right), \\ &\text{for all } t \geq t_0, \end{aligned}$$
(31)

and hence,

$$|X_{1}(t)| \leq \hat{\beta} (|X_{1}(t_{0})|, 0) + \hat{\gamma} \left( 2 \sup_{\phi(t_{0}) \leq \theta \leq t_{0}} |Z_{2}(\theta)| \right),$$
  
for all  $t \geq t_{0}.$  (32)

Setting  $s = \frac{t+t_0}{2}$  in (30) we have that

$$|X_{1}(t)| \leq \hat{\beta}\left(\left|X_{1}\left(\frac{t+t_{0}}{2}\right)\right|, \frac{t-t_{0}}{2}\right) + \hat{\gamma}\left(\sup_{\phi\left(\frac{t+t_{0}}{2}\right) \leq \theta \leq \phi(t)} |Z_{2}(\theta)|\right).$$
(33)

We estimate now  $\sup_{\phi\left(\frac{t+t_0}{2}\right) \le \theta \le \phi(t)} |Z_2(\theta)|$ . Solving (22) we get

$$\sup_{\substack{\phi\left(\frac{t+t_{0}}{2}\right) \le \theta \le \phi(t)}} |Z_{2}(\theta)|$$

$$\leq 2 \sup_{t_{0}-D(X_{1}(t_{0})) \le \theta \le t_{0}} |Z_{2}(\theta)| e^{-c_{2}\left(\phi\left(\frac{t+t_{0}}{2}\right)-t_{0}\right)},$$
for all  $t \ge 2\sigma(t_{0}) - t_{0}.$ 
(34)

With the help of relations (22) and (34) we get

$$\sup_{\substack{\phi\left(\frac{t+t_0}{2}\right) \le \theta \le \phi(t)}} |Z_2(\theta)| \le \sup_{\substack{\phi\left(\frac{t+t_0}{2}\right) \le \theta \le t_0}} |Z_2(\theta)| + \sup_{t_0 \le \theta \le \phi(t)} |Z_2(\theta)|$$
$$\le 2 \sup_{t_0 - D(X_1(t_0)) \le \theta \le t_0} |Z_2(\theta)|,$$
for all  $t_0 \le t \le 2\sigma(t_0) - t_0.$  (35)

Hence, using the fact that  $\phi\left(\frac{t+t_0}{2}\right) = \frac{t+t_0}{2} - D\left(X_1\left(\frac{t+t_0}{2}\right)\right)$  we get for all  $t \ge t_0$ 

$$\sup_{\substack{\{z_{1} \in U_{1} \\ z_{1} \in U_{2} \in U_{2} \\ z_{1} \in D(X_{1}(t_{0})) \le \theta \le t_{0} \\ x_{0} = C_{1}(D(0) + \delta_{1}(|x_{1}(\frac{t+t_{0}}{2})|))e^{-\frac{c_{2}}{2}(t-t_{0})}.$$
(36)

$$\Pi_{1}(\theta) = X_{1}(t) + \int_{t-D(X_{1}(t))}^{\theta} \frac{f_{1}(\sigma(s), \Pi_{1}(s), \kappa(\sigma(s), \Pi_{1}(s)) + Z_{2}(s)) \,\mathrm{d}s}{1 - \nabla D(\Pi_{1}(s)) f_{1}(\sigma(s), \Pi_{1}(s), \kappa(\sigma(s), \Pi_{1}(s)) + Z_{2}(s))}, \quad t - D(X_{1}(t)) \le \theta \le t.$$

$$(24)$$

Box I.

Setting  $s = t_0$  and replacing t by  $\frac{t+t_0}{2}$  we get from (30) that

$$\left| X_1\left(\frac{t+t_0}{2}\right) \right| \leq \hat{\beta} \left( |X_1(t_0)|, \frac{t-t_0}{2} \right) + \hat{\gamma} \left( \sup_{\phi(t_0) \leq \theta \leq \phi\left(\frac{t+t_0}{2}\right)} |Z_2(\theta)| \right).$$
(37)

Since,

$$\sup_{\phi(t_0) \le \theta \le \phi\left(\frac{t+t_0}{2}\right)} |Z_2(\theta)| \le \sup_{\phi(t_0) \le \theta \le t_0} |Z_2(\theta)| + \sup_{t_0 \le \theta \le \phi(t)} |Z_2(\theta)|$$

$$\leq 2 \sup_{t_0 - D(X_1(t_0)) \le \theta \le t_0} |Z_2(\theta)|, \qquad (38)$$

we arrive at

$$\left| X_1\left(\frac{t+t_0}{2}\right) \right| \leq \hat{\beta} \left( |X_1(t_0)|, 0 \right) + \hat{\gamma} \left( 2 \sup_{t_0 - D(X_1(t_0)) \leq \theta \leq t_0} |Z_2(\theta)| \right), \text{ for all } t \geq t_0.$$
(39)

Combining (29), (32), (33), (36) and (39), we get the statement of the lemma with

$$\beta^{*}(s, t - t_{0}) = \hat{\beta} \left( \hat{\beta} (s, 0) + \hat{\gamma} (2s), \frac{t - t_{0}}{2} \right) + \hat{\gamma} \left( 2se^{c_{2} \left( D(0) + \delta_{1} \left( \hat{\beta}(s, 0) + \hat{\gamma}(2s) \right) \right)} e^{-\frac{c_{2}}{2} (t - t_{0})} \right) + 2se^{c_{2} \left( D(0) + \delta_{1} \left( \hat{\beta}(s, 0) + \hat{\gamma}(2s) \right) \right)} e^{-c_{2} (t - t_{0})}. \quad \Box \qquad (40)$$

**Lemma 4** (Bound on Predictor in Terms of System's States). There exists a class  $\mathscr{KC}_{\infty}$  function  $\xi$  such that for all solutions of system (1), (2) satisfying (12) for 0 < c < 1, the following holds

$$|P_1(\theta)| \le \xi \left( |X_1(t)| + \sup_{t - D(X_1(t)) \le \tau \le t} |X_2(\tau)|, c \right),$$
  
$$t - D(X_1(t)) \le \theta \le t.$$
(41)

**Proof.** Consider the following ODE in  $\theta$  which follows by differentiating (5) together with the initial condition  $P_1(t - D(X_1(t))) = X_1(t)$ 

$$\frac{\mathrm{d}P_1(\theta)}{\mathrm{d}\theta} = \frac{f_1\left(\sigma\left(\theta\right), P_1\left(\theta\right), X_2\left(\theta\right)\right)}{1 - \nabla D\left(P_1\left(\theta\right)\right) f_1\left(\sigma\left(\theta\right), P_1\left(\theta\right), X_2\left(\theta\right)\right)},$$
  
$$t - D\left(X_1(t)\right) \le \theta \le t.$$
 (42)

With the change of variables

$$y = \sigma(\theta), \tag{43}$$

we re-write (42) as

$$\frac{dP_1(\phi(y))}{dy} = f_1(y, P_1(\phi(y)), X_2(y - D(P_1(\phi(y))))),$$
  
 $t \le y \le \sigma(t).$  (44)

Using (14) we get

$$\frac{\mathrm{d}R\left(y,P_{1}(\phi(y))\right)}{\mathrm{d}\theta}\frac{\mathrm{d}\theta}{\mathrm{d}y}$$

$$\leq R\left(y,P_{1}(\phi(y))\right) + \alpha_{3}\left(|X_{2}\left(y-D\left(P_{1}(\phi(y))\right)\right)|\right). \tag{45}$$

With (12) we have

$$\frac{dR(\sigma(\theta), P_{1}(\theta))}{d\theta} \leq \frac{1}{1-c} \left( R(\sigma(\theta), P_{1}(\theta)) + \alpha_{3}\left( |X_{2}(\theta)| \right) \right),$$

$$t - D(X_{1}(t)) \leq \theta \leq t.$$
(46)

Under Assumption 1, there exists a function  $\delta_1 \in \mathscr{K}_{\infty} \cap C^1$  such that

$$D(X_1) \le D(0) + \delta_1(|X_1|), \qquad (47)$$

and hence, using the comparison principle we have from (46) for all  $t - D(X_1(t)) \le \theta \le t$  that

$$R\left(\sigma\left(\theta\right), P_{1}\left(\theta\right)\right) \leq e^{\frac{D\left(0\right)+\delta_{1}\left(\left|X_{1}\left(t\right)\right|\right)}{1-c}} \left(R\left(t, X_{1}\left(t\right)\right)\right) + \sup_{t-D\left(X_{1}\left(t\right)\right)\leq\tau\leq t}\alpha_{3}\left(\left|X_{2}\left(\tau\right)\right|\right)\right).$$
(48)

With standard properties of class  $\mathscr{K}_{\infty}$  functions we get the statement of the lemma with  $\xi \in \mathscr{KC}_{\infty}$  given by

$$\xi(s,c) = \alpha_1^{-1} \left( (\alpha_2(s) + \alpha_3(s)) e^{\frac{D(0) + \delta_1(s)}{1 - c}} \right). \quad \Box$$
(49)

**Lemma 5** (Bound on Predictor in Terms of Transformed System's States). There exists a class  $\mathcal{K}$  function  $\gamma$  such that for all solutions of system (1), (2) satisfying (12) for 0 < c < 1, the following holds

$$|\Pi_{1}(\theta)| \leq \gamma \left( |X_{1}(t)| + \sup_{t - D(X_{1}(t)) \leq \tau \leq t} |Z_{2}(\tau)| \right),$$
  
$$t - D(X_{1}(t)) \leq \theta \leq t.$$
(50)

**Proof.** Let Y(s) be the solution of  $\frac{dY(s)}{ds} = f_1(s, Y(s), \kappa(s, Y(s)) + \omega(s))$  for  $s \ge s_0 \ge 0$ . Under Assumption 3 and [34], (see also [35,36]), there exist class  $\mathscr{KL}$  function  $\hat{\beta}$  and class  $\mathscr{K}$  function  $\hat{\gamma}$  such that

$$|Y(s)| \le \hat{\beta} (|Y(s_0)|, s - s_0) + \hat{\gamma} \left( \sup_{s_0 \le r \le s} |\omega(r)| \right),$$
  
for all  $s \ge s_0.$  (51)

Using the change of variable (43) and definition (24), we have that

$$\frac{d\Pi_{1}(\phi(y))}{dy} = f_{1}(y, \Pi_{1}(\phi(y)), \kappa(y, \Pi_{1}(\phi(y))) + Z_{2}(\phi(y))),$$
  
$$t \le y \le \sigma(t).$$
(52)

Since  $\Pi_1(\phi(y))$  satisfies the same ODE in y as the ODE for Y(s) in s we have that

$$|\Pi_1(\phi(y))| \le \hat{\beta} \left( |X_1(t)|, y-t \right) + \hat{\gamma} \left( \sup_{t \le y \le \sigma(t)} |Z_2(\phi(y))| \right),$$
  
for all  $t \le y \le \sigma(t)$ . (53)

Using (43) (which can be also written as  $\theta = \phi(y)$ ) with the fact that  $\hat{\beta}(s, r) \leq \hat{\beta}(s, 0)$  for all  $r \geq 0$ , we get from (53)

$$|\Pi_{1}(\theta)| \leq \hat{\beta} \left(|X_{1}(t)|, 0\right) + \hat{\gamma} \left( \sup_{t - D(X_{1}(t)) \leq \tau \leq t} |Z_{2}(\tau)| \right),$$
  
$$t - D\left(X_{1}(t)\right) \leq \theta \leq t.$$
(54)

With the properties of class  $\mathscr{K}$  functions we get the statement of the lemma where  $\gamma(s) = \hat{\beta}(s, 0) + \hat{\gamma}(s)$ .  $\Box$ 

**Lemma 6** (Equivalence of Norms for Original and Target System). There exist a function  $\xi_1$  of class  $\mathscr{KC}_{\infty}$  and a class  $\mathscr{K}_{\infty}$  function  $\alpha_4$  such that for all solutions of system (1), (2) satisfying (12) for 0 < c < 1, the following holds

$$\begin{aligned} |X_{1}(t)| + \sup_{t - D(X_{1}(t)) \le \theta \le t} |Z_{2}(\theta)| \\ \le \xi_{1} \left( |X_{1}(t)| + \sup_{t - D(X_{1}(t)) \le \theta \le t} |X_{2}(\theta)|, c \right) \\ |X_{1}(t)| + \sup_{t - D(X_{1}(t)) \le \theta \le t} |X_{2}(\theta)| \end{aligned}$$
(55)

$$\leq \alpha_4 \left( |X_1(t)| + \sup_{t - D(X_1(t)) \leq \theta \leq t} |Z_2(\theta)| \right), \tag{56}$$

for all  $t \geq t_0$ .

**Proof.** Using the direct backstepping transformation (20) and the bounds (15), (41) we get the bound (55) with  $\xi_1(s, c) = s + \hat{\rho}(\xi(s, c))$ . Using the inverse backstepping transformation (23) and the bounds (15), (50) we get the bound (56) with  $\alpha_4(s) = s + \hat{\rho}(\gamma(s))$ .  $\Box$ 

**Lemma 7** (Ball Around the Origin within the Feasibility Region). There exists a function  $\xi_c$  of class  $\mathscr{KC}_{\infty}$  such that for all solutions of system (1), (2) that satisfy

$$|X_1(t)| + \sup_{t - D(X_1(t)) \le \theta \le t} |X_2(\theta)| < \bar{\xi}_c(c, c),$$
(57)

for 0 < c < 1 also satisfy (12).

**Proof.** Since  $D \in C^1(\mathbb{R}^n; \mathbb{R}_+)$  there exists class  $\mathscr{K}_{\infty}$  function  $\delta_2$  such that

$$|\nabla D(X_1)| \le |\nabla D(0)| + \delta_2(|X_1|).$$
(58)

If a solution satisfies

$$(|\nabla D(0)| + \delta_2 (|P_1(\theta)|)) \hat{\alpha} \left( |P_1(\theta)| + \sup_{t - D(X_1(t)) \le \theta \le t} |X_2(\theta)| \right)$$
  
$$\leq c, \quad t - D(X_1(t)) \le \theta \le t,$$
(59)

for 0 < c < 1, then it also satisfies (12). Using Lemma 4, (59) is satisfied for 0 < c < 1 as long as bound (57) holds, where the class  $\mathscr{KC}_{\infty}$  function  $\xi_c$  is given by

$$\xi_c(s,c) = (|\nabla D_1(0)| + \delta_2(\xi(s,c))) \hat{\alpha}(\xi(s,c) + s),$$
(60)

and with  $\xi_c$  we denote the inverse function of  $\xi_c$  with respect to  $\xi_c$ 's first argument.  $\Box$ 



Fig. 2. A typical marine cooling system with one consumer.

**Lemma 8** (Estimate of the Region of Attraction). There exists a class  $\mathscr{K}$  function  $\xi_{\text{RoA}}$  such that for all initial conditions of the closed-loop system (1), (2), (4) and (5) that satisfy (16), the solutions of system (1), (2) satisfy (57) for 0 < c < 1 and hence satisfy (12).

**Proof.** Using Lemma 6, with the help of (25) we have that

$$\Omega(t) \le \alpha_4 \left( \beta^* \left( \xi_1 \left( \Omega(t_0), c \right), t - t_0 \right) \right), \tag{61}$$

where

$$\Omega(t) = |X_1(t)| + \sup_{t \to D(X_1(t)) \le \theta \le t} |X_2(\theta)|.$$
(62)

Hence, for all initial conditions that satisfy bound (16) with any class  $\mathscr{K}$  choice  $\xi_{\text{RoA}}(c) \leq \overline{\xi^*}_{\text{RoA}}(\overline{\xi}_c(c,c),c)$ , where  $\overline{\xi^*}_{\text{RoA}}(s,c)$  is the inverse of the function  $\xi^*_{\text{RoA}}(s,c) = \alpha_4 (\beta^* (\xi_1(s,c),0)) \in \mathscr{KC}_{\infty}$  with respect to  $\xi^*_{\text{RoA}}$ 's first argument, the solutions satisfy (57). Moreover, for all those initial conditions, the solutions verify (12) for all  $\theta \geq t_0 - D(X_1(t_0))$ .  $\Box$ 

**Proof of 1.** Using (61) we get (17) with  $\beta(s, t) = \alpha_4(\beta^*(s, t))$ . From (1), Assumption 1 and the Lipschitzness of  $X_2$  in  $[t_0 - D(X_1(t_0)), t_0)$  guarantee the existence and uniqueness of  $X_1 \in C^1[t_0, \sigma(t_0))$  where  $\sigma(t_0) = t_0 + D(X_1(\sigma(t_0)))$ . The target system (21) and (22) guarantees the existence and uniqueness of  $X_1 \in C^1(\sigma(t_0), \infty)$ , and (23) with the continuity of  $X_2$  at  $t_0$  guarantees that  $X_1 \in C^1[t_0, \infty)$ . Using the fact that  $\Pi_1$  satisfies  $\dot{\Pi}_1 = \frac{f_1(\sigma(t), \Pi_1(t), \kappa(\sigma(t), \Pi_1(t)) + Z_2(t))}{1 - \nabla D(\Pi_1(t)) f_1(\sigma(t), \Pi_1(t), \kappa(\sigma(t), \Pi_1(t)) + Z_2(t))}$  for  $t \ge t_0$ , the local Lipschitzness of  $\nabla D$ , relation (12) and Assumption 3 ( $\kappa \in C^1([t_0, \infty) \times \mathbb{R}^n; \mathbb{R})$ ) guarantee that  $\Pi_1 \in C^1(t_0, \infty)$ . Using (23), with the help of (22) and the fact that  $\kappa \in C^1([t_0, \infty) \times \mathbb{R}^n; \mathbb{R})$  we get that  $X_2 \in C^1(t_0, \infty)$ . Using Lemma 8 together with (47) and (58) we get (18), (19) with any class  $\mathscr{K}$  choice  $\delta^*(c) \ge \delta_1(\bar{\xi}_c(c, c))$ .

#### 4. Example 1: application to cooling systems

In marine transportation of materials, the design of control laws for the ship's cooling system is of paramount importance due to the significant potential of the cooling system in terms of energy optimization [2]. In Fig. 2, we show a typical marine cooling circuit with one consumer, denoted by C, and a Heat Exchanger. We denote with  $T_{in}$  the input temperature towards the consumer, i.e., the output temperature of the Heat Exchanger. Due to the transportation time of the coolant (typically water) from the Heat Exchanger to the consumer *C*, the actual input temperature  $T_{in}$  in the consumer is delayed by *H*, namely,  $T_{in}(t - H)$ . The delay time *H* depends on the flow rate  $q_2$  which can be controlled through a pump. In order to design a feedback law  $q_2$  we take into account that the flow rate  $q_2$  has to be proportional to the temperature at the other end of the consumer, which we denote with  $T_{out}$ . This is because it makes sense to increase the flow rate if the outer temperature of the consumer *C* is increasing. A simple choice is  $q_2 = k_1 T_{in} + k_2$ . The control objective is to regulate the temperature



Fig. 3. Response of the cooling model (63)-(64) with the predictor-based controller (65) (solid line) and in open-loop (dashed line).

 $T_{\text{out}}$  to a constant set-point, say  $T_{\text{eq}} > 0$ . This is achieved by controlling the flow rate  $q_1$  at the input of the Heat Exchanger through a pump.

Denoting by  $T_{out} = X_1$ ,  $T_{in} = X_2$ ,  $q_2 = k_1X_1 + k_2$ ,  $H = \frac{b}{q_2} = D$ and  $q_1 = U$  and by neglecting the effect of the hydraulics in the system (since the hydraulic dynamics assumed to be much faster than the heat dynamics [2]), the equations that describe the thermodynamics of the cooling circuit of Fig. 2 are

$$\dot{X}_{1}(t) = a \left( X_{1}(t) - X_{2} \left( t - \frac{b}{k_{1}X_{1}(t) + k_{2}} \right) \right) \\ \times (k_{1}X_{1}(t) + k_{2})$$
(63)

$$\dot{X}_2(t) = (k_1 X_1(t) + k_2) (X_1(t) - X_2(t)) - U(t),$$
(64)

where a < 0, and  $b, k_1, k_2 > 0$ . Since the coolant flows only in one direction  $q_1 > 0$ , and hence U > 0. Since the coolant is typically water, both  $T_{out}$  and  $T_{in}$  cannot fall below zero, and hence  $X_1, X_2 > 0$ . In addition since the consumer always adds heat (due to its functioning),  $T_{out} \ge T_{in}(t - H)$  and  $T_{out} \ge T_{in}$ , and hence  $X_1(t) - X_2\left(t - \frac{b}{k_1X_1+k_2}\right) \ge 0$  and  $X_1(t) - X_2(t) \ge 0$  for all  $t \ge 0$ . Assumption 2 is satisfied (in the domain of interest) since  $X_1$ remains bounded because  $q_2 = k_1X_1(t) + k_2 \ge 0$ , which follows form the fact that if  $q_2 = 0$  then  $\dot{X}_1(t) = 0$  (so  $q_2$  cannot cross from being positive to being negative). Assumption 1 is satisfied for all  $X_1 > 0$  and Assumption 3 is satisfied since for  $q_2 > 0$ , (63) is inputto-state stable from the "disturbance"  $w = X_2\left(t - \frac{b}{k_1X_1(t)+k_2}\right)$ . We choose the control law  $\kappa(X_1) = X_1 + \frac{c_1}{a} \frac{X_1}{k_1X_1+k_2}$ , and hence, the predictor-based control law for this system becomes

$$U(t) = (k_1 X_1(t) + k_2) (X_1(t) - X_2(t))$$

$$+ c_{2} \left( X_{2}(t) - P_{1}(t) - \frac{c_{1}}{a} \frac{P_{1}(t) - T_{eq}}{k_{1}P_{1}(t) + k_{2}} \right) - \left( 1 + \frac{c_{1}}{ak_{1}} \frac{T_{eq} + \frac{k_{2}}{k_{1}}}{\left(P_{1}(t) + \frac{k_{2}}{k_{1}}\right)^{2}} \right) \times \frac{a \left(P_{1}(t) - X_{2} \left(t\right)\right) \left(k_{1}P_{1}(t) + k_{2}\right)}{1 + \frac{bk_{1}}{\left(k_{1}P_{1}(t) + k_{2}\right)^{2}} a \left(P_{1}(t) - X_{2} \left(t\right)\right) \left(k_{1}P_{1}(t) + k_{2}\right)}, \quad (65)$$

where  $P_1(t)$  is defined in Box II. We choose the parameters of the plant and of the controller as a = -1 and  $c_1 = c_2 = b = k_1 = k_2 = 1$  and the initial conditions as  $X_1(0) = 1$  and  $X_2(\theta) = 0.2$  for all  $-\frac{b}{k_1X_1(0)+k_2} \le \theta \le 0$ . We show in Fig. 3 the temperatures  $T_{out}$ ,  $T_{in}$  together with the input flow  $q_1$ . We compare the response of the system with the predictor-based controller (65) and with no control. In the case of the open-loop response we observe that the temperatures  $T_{out}$ ,  $T_{in}$  converge to the same value but not at the desired set-point. This is what one expects since the system has an equilibrium at  $X_1 = X_2$ . In contrast, the predictor based controller regulates the temperatures  $T_{out}$ ,  $T_{in}$  at the desired set point  $T_{eq}$ .

#### 5. Example 2

We consider the system

$$\dot{s}(t) = v(t - r_1 \sin^2(\omega s(t)))$$
(67)

$$\dot{v}(t) = a(t),\tag{68}$$

where the state variables are denoted with s and v and the control variable is denoted with a(t). This system has some resemblance with the model considered in [8] for the "soft" automatic



**Fig. 4.** Response of system (67)–(68) with the controller (69)–(70) and initial conditions as s(0) = 1 and  $v(\theta) = 0.1$ , for all  $-r_1 \sin^2(\omega s(0)) \le \theta \le 0$ .

landing. In view of condition (12) one should expect that global stabilization is not achievable. This fact can be viewed in terms of the system dynamics (67)–(68) and in particular from relation  $v(t - r_1 \sin^2(\omega s(t)))$ . Subsystem (67) is forward-complete from v since it is linear, and the delay  $r_1 \sin^2(\omega s(t))$  satisfies Assumption 1. Hence, Theorem 1 applies. We choose the parameters of the plant as  $r_1 = 0.3$  and  $\omega = 15$ , the initial conditions of the plant as s(0) = 1 and  $v(\theta) = 0.1$  for all  $-r_1 \sin^2(\omega s(0)) \le \theta \le 0$  and the parameters of the nominal controller as  $c_1 = c_2 = 0.5$ . The predictor-based controller is given by

$$a(t) = -c_2 (v(t) + c_1 P_1(t)) - c_1 \frac{v(t)}{1 - r_1 \omega \sin(\omega P_1(t)) \cos(\omega P_1(t)) v(t)},$$
(69)

where for all  $t - r_1 \sin^2(\omega s(t)) \le \theta \le t$ 

$$P_{1}(\theta) = s(t) + \int_{t-r_{1}\sin^{2}(\omega s(t))}^{\theta} \frac{v(s) \, ds}{1 - r_{1}\omega\sin(\omega P_{1}(s))\cos(\omega P_{1}(s))v(s)}.$$
 (70)

The control signal reaches the state *s* at the time  $t^*$  which satisfies  $0.3 \sin^2 (15(0.1t^* + 1)) = t^* = 0.0887$ . In Fig. 4, we show the response of the system. As Theorem 1 predicts s(t), v(t) converge to zero, whereas  $\phi(t)$  and  $\sigma(t)$  remain increasing for all times. From Fig. 4, we also observe that at the time instants where  $0.3 \sin^2(15s(t_1)) = 0$  we have that  $\phi(t) = t = \sigma(t)$ . Moreover,

the peaks of the control signal a(t) occur at the time instants where  $\sigma(t)$  increases rapidly.

# 6. Conclusions

Although we consider plants with only state-dependent state delay the results of this paper can be extended to the case of simultaneous state-dependent input and state delays. The tools that one has to use are the ones from the present paper and from [31]. However, the stability analysis will be much more involved: one has to satisfy not only one, but two (one for each delay) feasibility conditions.

Since we obtain only regional results one may wonder if the results of this paper can be applied to locally stabilizable plants. The answer to this question is positive by applying similar techniques to the ones from [32].

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