Time- and State-Dependent Input Delay-Compensated Bang-Bang Control of a Screw Extruder for 3D Printing

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Abstract—In this paper a delay-compensated Bang-Bang control design methodology for the control of the nozzle output flow rate of screw-extruder-based 3D printing processes is developed. The presented application has a great potential to move beyond the most commonly used processes such as Fused Deposition Modeling (FDM) and Syringe Based Extrusion (SBE), improving the build speed and the 3D parts accuracy. A geometrical decomposition of the screw extruder in a partially and a fully filled regions (PFZ and FFZ) allows to describe the material convection in the extruder chamber by a 1D hyperbolic Partial Differential Equation (PDE) coupled with an Ordinary Differential Equation (ODE). After solving the hyperbolic PDE by the Method of Characteristics (MC), the coupled PDE-ODE’s system is transformed into a nonlinear state-dependent input delay system. The aforementioned delay system is extended to the non-thermal case with the consideration of periodic fluctuations acting on the material’s convection speed, which represent the effect of viscosity variations due to temperature changes in the extruder chamber, resulting to a nonlinear system with an input delay that simultaneously depends on the state and the time variable. Global Exponential Stability (GES) of the nonlinear delay-free plant is established under a piecewise exponential feedback controller that is designed. By combining the nominal, piecewise exponential feedback controller with nonlinear predictor feedback the compensation of the time- and state-dependent input delay of the extruder model is achieved. Global Asymptotic Stability (GAS) of the closed-loop system under the Bang-Bang predictor feedback control law is established when certain conditions, which are easy to verify, related to the extruder design and the material properties, as well as to the magnitude and frequency of the materials transport speed variations, are satisfied. Several simulations results are presented to illustrate the effectiveness of the proposed control design.

I. INTRODUCTION

Additive Manufacturing (AM) has a promising future and demonstrates its effectiveness in various applications involving tissue engineering [1], [2], chemical engineering [3], thermoplastics [4], metal [5] and ceramic [6] material’s fabrication. Functional 3D objects with complex geometrical shape can be produced in a short time without the need of tools thanks to the Computer Aid Design (CAD) that drastically reduces the products development procedure. Currently, the most popular plastics 3D printers are based on FDM [7]–[9] and SBE [4], [10] technologies (Fig.1).

In these processes, biodegradable polymers are transported, heated and pressurized in an extruder chamber before being dropped on a platform, one horizontal thin layer at a time, until the complete 3D part is built such that it closely resembles the original CAD model. One of the crucial point that is not commonly addressed in the existing literature of extrusion-based 3D printing is controlling of the start and stop of extrusion-on-demand. A hybrid extrusion force-velocity modeling and tracking control for the fabrication of functionally graded material parts is developed in [11] using a first order differential equation that describes the plunger dynamic in a SBE process. Some extents of that approach are proposed by [12] with a robust tracking of the extrusion force to recover constant flow disturbances whereas [13] considers an unknown transfer function gain with an adaptive control strategy. Several issues regarding on FDM are discussed in [14], [15] and references therein, including the potential clogging due to agglomerate formation at the nozzle, appearance of bubbles, density inhomogeneity, tracking of short time-scale process variations, and prediction of anomalies such as material overflow and underflow for diverse applications. Thermal control is left out of most prior studies which are essentially based on empirical models.

![Fig. 1. (a) FDM, (b) SBE, (c) SE processes [4].](image)

In this paper, we are interested in the flow control issues related to the recent advances of 3D printing technology for which a Screw Extrusion (SE) process is utilized. In SE, the rotating screw allows a continuous feeding mechanism and generates a sufficiently high pressure in the extruder chamber, increasing, as a result, the printing speed. In addition, the screw motion extends the mixing capabilities of the system, and thereby, reduces drastically the risk of potential clogging...
at the nozzle while improving the homogeneity of the extruded filament [4], [16]. The SE process with granular material moves beyond the restrictions of FDM and does not require filament-shaped raw materials to operate. Consequently, it enables the processing of a broader range of raw materials and permits an easy recycling of wasted plastic during extrusion [4], [16]. In such processes, the need to control the start and stop of the extrusion process on demand calls for advanced control methodologies that are capable of enhancing the final product’s quality in an industrial level. Even if experimental results demonstrate the effectiveness of SE [16], [4], the challenging control problems arising in such applications are actually poorly investigated.

In the present article, a generic and dynamical model of a homogeneous melt SE process derived from mass and momentum balance laws [17], [18] is used for the design of a delay-compensated Bang-Bang controller which permits a fast and accurate control of the flow at the nozzle output. The model consists of a 1D Partial Differential Equation (PDE) that is defined on a time-varying spatial domain whose dynamics obey to an Ordinary Differential Equation (ODE). The transformation of the coupled PDE-ODE system into a state-dependent input delay system, which describes the dynamics of the material convection in the extruder chamber, is achieved after solving the PDE by the Method of Characteristics (MC) [17], [19]. In order to also account for potential periodic fluctuations of the materials transport speed when processing granular pellets [21], due to the thermal energy that is supplied into the system from the heater of the extruder and due to the mechanical shearing effect by the rotation of the screw, the state-dependent input delay model is extended to a nonlinear system with an input delay that depends simultaneously on the state and the time variable (see [20], [23] and [22] for the treatment of systems with time- and state-dependent delays).

In [19], a delay-compensated Bang-Bang control law is developed for the control of the nozzle output flow rate of an isothermal screw extrusion process, achieving GES of the delay-free plant at any given setpoint. By combining the nominal, piecewise exponential feedback controller [19] with nonlinear predictor feedback, which is extended from the state-dependent input delay case [20] to the case in which the vector field and the delay function depend explicitly on time, the compensation of the time- and state-dependent input delay of the non-isothermal screw extrusion model is achieved. GAS of the closed-loop system under the delay-compensated Bang-Bang controller is established when certain conditions, related to the extruder design and the material properties, as well as to the periodic fluctuations, are satisfied. Several simulations results are presented including the case in which there is uncertainty in the value of the periodic variations of the material’s transport speed.

This paper is organized as follows: The screw extruder mechanisms and the bi-zone model of the extruder consisting of the transport PDE coupled with the ODE for the moving interface is discussed in Section [II]. In Section [III] the transformation of the coupled PDE-ODE system into a state-dependent input delay system by computing the PDEs solution by the MC is presented and it is then extended to a nonlinear system with a time- and state-dependent input delay. The control of the delay-free plant with a piecewise exponential Bang-Bang-like control law is described in Section [IV]. In Section [V] we design the predictor feedback control law for nonlinear systems with time- and state-dependent delay acting on the input. The application of the predictor feedback control law to the screw extruder model is presented in Section [VI]. The paper ends with simulations, including a discussion on the robustness properties of a state-dependent input delay compensator to time- varying perturbations acting on the vector field and the delay function, in Section [VII].

II. 3D PRINTING BASED ON SINGLE-SCREW EXTRUDERS

A. Extrusion process description and structural decomposition of the extruder into a partially and a fully filled zone

A screw extruder is divided into one or several conveying zones (transport zones), melting zones (for material fusion) and mixing zones in which the extruded melt is submitted to high pressure, before its eviction through the nozzle [17], [18], [24]–[50]. The net flow rate at the extruder nozzle depends mainly on the material flow in the longitudinal direction given by 1D heat and mass transport equations [31], [32]. Another particularity of these processes is that they can be divided in geometric regions which are partially and fully filled called PFZ and FFZ, respectively (Fig. 2). The PFZ which is submitted to an atmospheric pressure is a conveying region and the flow in the FFZ is determined by the pressure gradient building-up in that region due to the nozzle resistance. These two zones are coupled by an interface which moves according to the volume of material accumulated in the FFZ. Basically, the moving interface is located at the point where the pressure gradient passes from zero to a non null value.

![Fig. 2. Bi-zone model of a screw extruder.](image)

B. Mass and momentum balance of an extrusion process

1) Mass balance of the PFZ: The PFZ is defined on the time-varying spatial interval \((x(t), L)\), \(x(t)\), being the length of the FFZ and \(L\) the extruder length (Fig. 2). The mass balance in this area can be expressed using the fraction of the effective volume between a screw element and the barrel \((V_{eff})\) which is occupied by the extruded material, namely, the filling
ratio \( u \). Considering an incompressible homogeneous mixture with constant density \( \rho_0 \) and viscosity \( \eta \), the following mass conservation equation is deduced

\[
\partial_t u(z, t) = \xi \frac{\partial_z u(z, t)}{\partial z}, \quad (t, z) \in (\mathbb{R}^+, (x(t), L))
\]

\[
u(L, t) = U(t),
\]

where \( N_0 \) is the constant screw speed and \( \xi \) the uniform pitch of the screw. The boundary condition \( u(L, t) \) is defined assuming the continuity of the flow at the inlet \( z = L \)

\[
U(t) = \frac{F_{in}(t)}{\rho_0 N_0 V_{ref}},
\]

where \( F_{in}(t) \) is the feeding rate. Physically, the term \( \rho_0 N_0 V_{ref} \) in (3) is the maximum pumping capacity of the screw.

2) Momentum balance of the FFZ: The FFZ whose filling ratio is equal to one is defined on the spatial domain \((0, x(t))\), where the coordinate \( z = 0 \) is the extruder’s end. The FFZ flow depends on the pressure gradient that appears in this region, resulting to backward or forward flow. The momentum balance which is derived from Navier-Stokes equations under stationary conditions yields the pressure gradient

\[
\partial_z P(z, t) = -\frac{\eta}{B} F_d(t),
\]

for all \((t, z) \in (\mathbb{R}^+, (0, x(t)))\), where \( B \) is a coefficient of pressure flow. The net flow rate \( F_d(t) \), in the case of a Poiseuille flow is expressed with the help of the nozzle conductance \( K_d \), the viscosity \( \eta \), and the pressure at the nozzle \( P(0, t) \) as

\[
\begin{align*}
F_d(t) &= \frac{K_d}{\eta} \Delta P(t), \\
\Delta P(t) &= P(0, t) - P_0.
\end{align*}
\]

3) Mass balance of the FFZ : The FFZ mass balance leads to an ODE which describes the time evolution of its length. This length denoted by \( x(t) \) determines the location of the small transfer region that is assimilated to the point at which the pressure changes from the atmospheric pressure \( P_0 \) to a different value \([18, 27, 28, 30, 33]\)

\[
\frac{dx(t)}{dt} = \frac{\rho_0 \xi N_0 V_{rec} u(x(t), t) - F_d(t)}{\rho_0 S_{ref} (1 - u(x(t), t))},
\]

where \( S_{ref} \) is the available section and \( V_{rec} = \xi S_{ref} \).

4) Coupling condition at the PFZ-FFZ interface : The coupling condition is imposing the pressure continuity at the spatial coordinate \( x(t) \)

\[
P(x^-, t) = P(x^+, t) = P_0.
\]

Integrating the pressure gradient equation (4), the net flow rate defined in (5) is written as

\[
F_d(t) = \frac{K_d V_{rec} N_0 \rho_0 x(t)}{B \rho_0 + K_d x(t)}.
\]

Substituting (5) in (6) and using (8), equation (6) for the length of the FFZ is written as

\[
\frac{dx(t)}{dt} = -\xi N_0 \frac{K_d x(t)}{(B \rho_0 + K_d x(t))} \frac{u(x(t), t)}{(1 - u(x(t), t))}.
\]

III. FROM MASS BALANCE EQUATIONS OF THE EXTRUDER TO A DELAY SYSTEM

A. Isothermal delay system model

The bi-zone model [1, 2, 9] can be reduced to a nonlinear state dependent input delay system [20]. The characteristic solutions of [1] with respect to the boundary condition (2) are

\[
u(z, t) = U \left( t - \frac{L - z}{\xi N_0} \right),
\]

Substituting (10) into (9), we derive the following nonlinear system

\[
\dot{x}(t) = \frac{K_d x(t)}{(B \rho_0 + K_d x(t))} \frac{1}{1 - U (t - D_s(x(t)))} \left(1 - \frac{\theta_2 x(t)}{U (t - D_s(x(t)))} \right), \quad U(t) \in [0, 1)
\]

The state-dependent input delay function is denoted as

\[
D_s(x(t)) = \frac{L - x(t)}{\xi N_0}.
\]

A detailed derivation of the ODE (11) and an extensive description of the screw-extruder model for 3D printing is given in [19].

B. Delay system representation in a non-isothermal case

In this section, we propose the following extension of the state-dependent delay model (11) to account for the viscosity changes due to the temperature variations occurring in the physical system when processing plastic pellets

\[
\dot{x}(t) = c(t) - \frac{\theta_2 x(t)}{(1 + \theta_2 x(t)) \left(1 - \frac{\theta_2 x(t)}{U(t - D_s(x(t)))} \right)}, \quad U(t) \in [0, 1),
\]

where

\[
\theta_1 = \xi N_0, \quad \theta_2 = \frac{K_d}{B \rho_0},
\]

and

\[
D(t, x(t)) = \frac{L - x(t)}{c(t)};
\]

\[
c(t) = \theta_1 (1 + \epsilon \cos(\omega t)),
\]

where, \( \epsilon < 1 \) is a positive constant and \( \omega \) is the mean value of the angular frequency of the periodic fluctuations. Our choice for the non-isothermal model (13) is motivated by the fact that the expansion of granular material into a plastic state due to the thermal effect leads implicitly to periodic fluctuations of the convection speed, namely, \( \theta_1 \) due to the viscosity variations [21]. Moreover, some nozzle “instabilities” phenomena may appear as short periodic distortions of the extrudate, due to the viscoelastic properties of the fluid, with magnitude smaller than one [21].
IV. CONTROL OF THE DELAY-FREE SYSTEM WITH A “BANG-BANG” CONTROL LAW

A. Open-loop stability

The starting point of the delay system controller design consists of the construction of a nonlinear control law that stabilizes the delay-free system

$$\dot{x}(t) = c(t) \left[ -\frac{\theta_2 x(t)}{(1 + \theta_2 x(t))(1 - U(t))} + \frac{U(t)}{(1 - U(t))} \right], \quad U(t) \in [0, 1]. \quad (18)$$

For $\epsilon < 1$, the time-varying speed of the material transport $c(t)$ is strictly positive and the open-loop stabilizing control law of the delay-free plant (18) is given by

$$v(x^*) = \frac{\theta_2 x^*}{1 + \theta_2 x^*}, \quad \forall x^* \in [0, L], \quad (19)$$

for the physical parameters of the extruder satisfying $\theta_2 L < 1$. This statement is directly derived considering the Lyapunov function $V = |c(t)|$, where $c(t) = x(t) - x^*$.

B. “Bang-Bang” controller design with piecewise exponential functions

For the feedback stabilization of (18), we consider two exponential functions (19):

- For $x(t) \leq x^*$, a left-exponential function

$$v_l(x, x^*) = v(x^*) + (v_{\text{max}} - v(x^*)) \frac{1 - e^{a_l(x^*)(x-x^*)}}{1 - e^{-a_l(x^*)(x-x^*)}}, \quad (20)$$

where $a_l(x^*) > 0$ is the gain of the left exponential control law. The function (20) takes values in $[v(x^*), v_{\text{max}}]$, where $v_{\text{max}} < 1$ is the maximal value of the inlet filling ratio, namely the maximal feeding capacity of the extruder. Therefore, $v_l(0) = v_{\text{max}}$, allows to set the inlet flow at its maximum capacity for a rapid refill action when the extruder is empty.

- For $x(t) \geq x^*$, a right-exponential function

$$v_r(x, x^*) = v(x^*) - v(x^*) \frac{1 - e^{-a_r(x^*)(x-x^*)}}{1 - e^{-a_r(x^*)(L-x^*)}} \quad (21)$$

where $a_r(x^*) > 0$ is the gain of the right exponential control law. The function (21) belongs into the interval $[0, v(x^*)]$ and the control action stops radically the flow when the extruder is completely filled, namely, $v_r(L) = 0$.

C. Extension of the “Bang-Bang” control law on the whole domain $(0, L)$

Next, we introduce the characteristic function of the domains $[0, x^*]$ and $[x^*, L]$ and write the extended control law as

$$v(x, x^*) = v_l(x, x^*) h(x^* - x) + v_r(x, x^*) h(x - x^*), \quad (22)$$

where $h$ is the Heaviside function.

A continuous slope function at the setpoint $x^*$ denoted by $S(x^*)$ is imposed to extend the left and the right exponential controllers (20) and (21), respectively into the differentiable piecewise exponential feedback law (22). The slope function is defined as $S(x) = -\frac{dv(x, x^*)}{dx}$ (the minus sign is conventional). More precisely, the key point of the design is to define a free parameter that may be specified by the user as the value of slope function at the equilibrium $S(x^*)$, under some restrictions that will be emphasize in this section. It is clear that, equating the assigned value $S(x^*)$ to both left and right slope functions of (20) and (21), we can easily derive the following relations

$$S(x^*) = \frac{a_l(x^*)(v_{\text{max}} - v(x^*))}{1 - e^{-a_l(x^*)(x-x^*)}}, \quad (23)$$

$$S(x^*) = \frac{a_r(x^*)v(x^*)}{1 - e^{-a_r(x^*)(L-x^*)}}. \quad (24)$$

The equations (23) and (24) are both transcendental and admit numerical solutions namely the suitable exponents parameters needed to the left and to the right of the setpoint to achieve the differentiability of the controller (22). These solutions $a_l(x^*) > 0$ (respectively, $a_r(x^*) > 0$) exist if the linear and exponential functions of $a_l(x^*)$ (respectively, $a_r(x^*)$) have a strictly positive intersection. Consequently, the desired slope function $S(x^*)$ should be above some minimum value denoted $S_{\text{min}}(x^*)$, for any given equilibrium in the physical domain $(0, L)$. More precisely, the two equations in (23) have strictly positive solutions if at the origin ($a_l = 0$ and $a_r = 0$), the slope of their linear part is less than the slope of their exponential part respectively.

- For the left exponential slope function (23), we define the linear and the exponential functions

$$\begin{align*}
\psi_l(a_l(x^*)) &= a_l(x^*)(v_{\text{max}} - v(x^*)) \\
\phi_l(a_l(x^*)) &= S(x^*)(1 - e^{-a_l(x^*)(x-x^*)}),
\end{align*} \quad (25)$$

a solution of (23) should satisfy

$$\begin{align*}
\psi_l(a_l) &= a_l(v_{\text{max}} - v(x^*)) \\
\frac{da_l}{da_l} &= \frac{da_l}{a_l}, \quad < 0.
\end{align*} \quad (26)$$

It follows that for all $x \in (0, x^*)$, $S(x^*)$ should satisfy the inequality

$$S(x^*) > \frac{v_{\text{max}} - v(x^*)}{x^*}. \quad (27)$$

- Decomposing the right exponential slope function (24) into

$$\begin{align*}
\psi_r(a_r(x^*)) &= a_r(x^*)v(x^*) \\
\phi_r(a_r(x^*)) &= S(x^*)(1 - e^{-a_r(x^*)(L-x^*)}),
\end{align*} \quad (28)$$

we deduce that a solution of (24) should satisfy:

$$\begin{align*}
\psi_r(a_r(x^*)) &= a_r(x^*)v(x^*) \\
\frac{da_r}{da_r} &= \frac{da_r}{a_r}, \quad < 0.
\end{align*} \quad (29)$$

Hence, for all $x \in (x^*, L)$

$$S(x^*) > \frac{v(x^*)}{L-x^*} \quad (30)$$
Finally, the minimal value of the setpoint slope \( S_{\text{min}}(x^*) \) above which the gains \( a_1(x^*) \) and \( a_2(x^*) \) ensure the differentiability of the extended control law \([22]\) on \((0, L)\) is given by

\[
S_{\text{min}}(x^*) = \frac{x^*}{1 + \frac{x^*}{\theta_2 x^*}} + 1 \max \left\{ \frac{v_{\text{max}} \left(1 + \frac{1}{\theta_2 x^*}\right) - 1}{x^*} ; 1 - \frac{L - x^*}{L - x^*} \right\}
\]  

(31)

The speed of the controller or its “aggressiveness” increases with the rise of the setpoint slope \( S(x^*) \). As it is illustrated in Fig. 3 with the characteristics of the control law for the setpoints \( x^* = 0.02 m \) and \( x^* = 0.16 m \) with different values of the setpoint slope value \( S(x^*) \).

![Graph showing the control function for different setpoint slopes](image)

**Theorem 1.** For any setpoint \( x^* \in [0, L] \) and for any chosen setpoint slope \( S(x^*) \in \mathbb{R} \) satisfying \( S(x^*) \geq S_{\text{min}}(x^*) \), where \( S_{\text{min}}(x^*) \) is given by \([31]\), taking the control gains \((a_1(x^*), a_2(x^*))\) as solutions of

\[
a_1(x^*)(v_{\text{max}} - v(x^*)) - S(x^*)(1 - e^{-a_1(x^*)}) = 0,
\]

(32)

\[
a_2(x^*)v(x^*) - S(x^*)(1 - e^{-a_2(x^*)(L-x^*)}) = 0.
\]

(33)

The closed-loop system consisting of \([18]\) with an initial condition \( x_0 \in [0, L] \) and the extended control law \([20] - [22]\) is GES at \( x = x^* \).

**Proof:** We rewrite the delay-free plant \([18]\) as

\[
\dot{e}(t) = \frac{c(t)(U(t) - v(x^*))}{(1 - v(x^*))(1 + \theta_2(e(t) + x^*))} - \frac{c(t)\theta_2 e(t)}{1 + \theta_2(e(t) + x^*)},
\]

(34)

where, \( e(t) = x(t) - x^* \). The control law \([22]\) is written as

\[
v(x(t), x^*) = v(e(t) + x^*, x^*).
\]

(35)

The extended control law \([22]\) is a decreasing function of \( x \) and consequently for all \( x(t) \in [0, L] \),

\[
\text{sgn}(v(e(t) + x^*, x^*) - v(x^*, x^*)) = -\text{sgn}(e(t)).
\]

(36)

Moreover, \( 0 \leq v(x, x^*) \leq v_{\text{max}} < 1 \) and \( v(x^*, x^*) = v(x^*) \) is defined as the setpoint open-loop control \([19]\). Next, we introduce the following Lyapunov function

\[
V = |e(t)|.
\]

(37)

Hence,

\[
\dot{V} = \dot{e}(t) \text{sgn}(e(t)),
\]

(38)

and with the help of \([36]\), by choosing \( U(t) = v(x(t), x^*) \), \([38]\) is written as

\[
\dot{V} = -\alpha(t) V - \beta(t),
\]

(39)

where \( \alpha(t) > 0 \) and \( \beta(t) \geq 0 \) for all \( x \in [0, L] \) and \( U \in [0, 1] \). The functions \( \alpha(t) \) and \( \beta(t) \) are given by

\[
\alpha(t) = \frac{c(t)\theta_2}{1 + \theta_2(e(t) + x^*)},
\]

(40)

\[
\beta(t) = \frac{1}{(1 - v(e(t) + x^*, x^*))}
\]

\[
\times \frac{c(t) |v(e(t) + x^*, x^*) - v(x^*)|}{(1 - v(x^*)) (1 + \theta_2(e(t) + x^*))},
\]

(41)

where, \( c(t) \geq \theta_1 (1 - e) \), for all \( t \geq 0 \). Therefore

\[
\dot{V} \leq -\frac{\theta_1 (1 - e)}{1 - \theta_2 L} V.
\]

(42)

From \([42]\) the closed-loop system is exponentially stable at \( x^* \in [0, L] \) for all \( x_0 \in [0, L] \).

**V. PREDICTOR FEEDBACK CONTROL FOR NONLINEAR SYSTEMS WITH TIME- AND STATE-DEPENDENT INPUT DELAY**

**A. Predictor feedback design**

We consider the following nonlinear system with a time- and state-dependent input delay

\[
\dot{x}(t) = f(t, x(t), U(\phi(t)))
\]

(43)

\[
\phi(t) = t - D(t, x(t)),
\]

(44)

where \( x \in \mathbb{R}^n \), \( U : [0 - D(t_0, x(t_0)), \infty) \rightarrow \mathbb{R} \), \( t \geq t_0 \geq 0 \), \( D \in C^1(\mathbb{R}^+ \times \mathbb{R}^n; \mathbb{R}^+ \), and \( f : \mathbb{R}^+ \times \mathbb{R}^n \times \mathbb{R} \rightarrow \mathbb{R}^n \) is locally Lipschitz with \( f(t, 0, 0) = 0 \) for all \( t \geq 0 \) and there exists a class \( K_\infty \) function \( \alpha \) such that

\[
|f(t, x, U)| \leq \alpha(|x| + |U|).
\]

(45)

The predictor feedback control law for system \([43] - [44]\) is

\[
U(t) = \kappa(\sigma(t), P(t)),
\]

(46)

where, for all \( t - D(t, x(t)) \leq \theta \leq t \)

\[
P(\theta) = x(t) + \int_{t-D(t,x(t))}^{\theta} \frac{f(\sigma(s), P(s), U(s))}{1 - F(\sigma(s), P(s), U(s))} ds,
\]

(47)

\[
\sigma(\theta) = t + \int_{t-D(t,x(t))}^{\theta} \frac{1}{1 - F(\sigma(s), P(s), U(s))} ds,
\]

(48)

and

\[
F(\sigma(\theta), P(\theta), U(\theta)) = \frac{\partial D}{\partial t} (\sigma(\theta), P(\theta))
\]

\[
+ \frac{\partial D}{\partial x} (\sigma(\theta), P(\theta)) \times f(\sigma(\theta), P(\theta), U(\theta)).
\]

(49)

When simulating the predictor feedback controller \([47] - [46]\), at each time step the ODE for the system \([43] - [44]\)
must be solved (using, for example, a simple Euler scheme) and the length of the delay must be computed (for example as the integer part of \( N(\tau) = \frac{D(t,x(t))}{\tau} \)), say \( N(\tau) \), where \( \tau \) is the discretization step. The predictor is then computed by integrating simultaneously the two integral relations (47) and (48) at each time step, using a numerical integration scheme. For instance, with the left endpoint rule of integration we get

\[
P(i) = x(i) + \tau \sum_{k=i-N(i)}^{k=i-1} f(\sigma(k), P(k), U(k)) \frac{1}{1 - F(\sigma(k), P(k), U(k))},
\]

where the prediction time is defined as

\[
\sigma(i) = i + \tau \sum_{k=i-N(i)}^{k=i-1} \frac{1}{(1 - F(\sigma(k), P(k), U(k))}.
\]

The prediction of the state at the time when the current control will have an effect on the state is defined as

\[
P(t) = x(t + D(\sigma(t), P(t))),
\]

where the prediction time is defined as

\[
\sigma(t) = t + D(\sigma(t), P(t)),
\]

which is derived from the inversion of the time variable \( t \rightarrow t - D(t, X(t)) \) in \( t \rightarrow t + D(\sigma(t), P(t)) \). Differentiating (52), (53) and using (43) we arrive at

\[
\frac{dx(\sigma(t))}{dt} = f(\sigma(t), x(\sigma(t)), U(t)) \frac{d\sigma(t)}{dt},
\]

and

\[
\dot{\sigma}(t) = \frac{1}{1 - F(\sigma(t), P(t), U(t))}.
\]

where \( F \) is defined in (49). Finally, the implicit integral relations (47) and (48) are derived by integrating (54) and (55) on the delay interval \([\phi(t), \theta] \).

The key point of the predictor feedback design is the feasibility condition defined as

\[
F_c := \frac{\partial D}{\partial t}(\sigma(\theta), P(\theta)) + \frac{\partial D}{\partial x}(\sigma(\theta), P(\theta)) f(\sigma(\theta), P(\theta), U(\theta)) < c,
\]

for all \( \theta \geq t_0 - D(t_0, x(t_0)) \) and some \( c \in (0, 1) \). Condition (56) guarantees that the feedback control action can reach the plant, namely, the delay rate is bounded by unity, and that the denominator of the predictor (47) and the prediction time (48) is positive. We refer the reader to [20] for details on the predictor feedback control design and analysis for systems with state-dependent input delay.

**B. Stability analysis**

**Assumption 1.** There exist a smooth positive definite function \( R \) and class \( K_\infty \) functions \( \mu_1, \mu_2 \) and \( \mu_3 \) such that for the plant \( \dot{x} = f(t, x, \omega) \), the following hold

\[
\mu_1(|x|) \leq R(t, x) \leq \mu_2(|x|)
\]

\[
\frac{\partial R(t, x)}{\partial t} + \frac{\partial R(t, x)}{\partial x} f(t, x, \omega) \leq R(t, x) + \mu_3(|\omega|),
\]

for all \((x, \omega) \in \mathbb{R}^{n+1} \) and \( t \geq t_0 \).

Assumption 1 guarantees that system \( \dot{x} = f(t, x, \omega) \) is strongly forward complete with respect to \( \omega \).

**Assumption 2.** There exist a locally Lipschitz function \( \kappa \in ([t_0, \infty) \times \mathbb{R}^n; \mathbb{R}) \) and a function \( \hat{\rho} \in K_\infty \) such that the plant \( \dot{x} = f(t, x(t), \kappa(t, x(t)) + \omega(t)) \) is input to state stable with respect to \( \omega \) and \( \kappa \) is uniformly bounded with respect to its first argument, that is,

\[
|\kappa(t, x)| \leq \hat{\rho}(|x|) \quad \forall \ t \geq t_0.
\]

**Assumption 3.** \( D \in C^1(\mathbb{R}_+ \times \mathbb{R}^n; \mathbb{R}_+) \), \( \frac{\partial D}{\partial \sigma} \) and \( \frac{\partial D}{\partial x} \) are locally Lipschitz (to guarantee the uniqueness of solutions), and there exist class \( K_\infty \) functions \( \mu_4, \mu_5, \mu_6 \) and non-negative constants \( c_1, c_2, c_3 \), with \( c_3 < c \), for some \( 0 < c < 1 \), such that

\[
D(t, x(t)) \leq c_1 + \mu_4(|x|),
\]

\[
\left| \frac{\partial D}{\partial x}(t, x(t)) \right| \leq c_2 + \mu_6(|x|),
\]

\[
\left| \frac{\partial D}{\partial \sigma}(t, x(t)) \right| \leq c_3 + \mu_5(|x|).
\]

The definitions of strong forward completeness and input-to-state stability are those from [34], and [35], respectively.

**Theorem 2.** Consider the closed-loop system consisting of the plant (43) and the control law (47). Under Assumptions 1 and 2 there exist a class \( K \) function \( \psi_{R_0A} \) and a class \( K_\infty \) function \( \beta_* \) such that for all initial conditions for which \( U \) is locally Lipschitz on the interval \([t_0 - D(t_0, x(t_0)), t_0] \) and which satisfy

\[
\Omega(t_0) < \psi_{R_0A}(c - c_3),
\]

for some \( 0 < c < 1 \), where

\[
\Omega(t) = |x(t)| + \sup_{t_0 - D(t, x(t)) \leq t} |U(\theta)|,
\]

there exists a unique solution to the closed-loop system with \( x \) Lipschitz on \([t_0, \infty) \), \( U \) Lipschitz on \((t_0, \infty) \), and the following holds

\[
\Omega(t) \leq \beta_*(\Omega(t_0), t - t_0),
\]

for all \( t \geq t_0 \). Furthermore, there exists a positive constant \( \gamma \) such that for all \( t \geq t_0 \),

\[
D(t, x(t)) \leq \gamma
\]

\[
\left| \frac{dD(t, x(t))}{dt} \right| \leq c
\]

**Proof of Theorem 2.** Estimates (65), (66), and (67) follow by directly applying Lemmas 1–8 from [20] (see the Appendix). Existence and uniqueness of a solution \( x \) Lipschitz on \([0, \infty) \) follows from the proof of Theorem 1 in [20] (page 7). It remains to show that \( U \) is Lipschitz on \([t_0, \infty) \). Since
\[ U(t) = \kappa (\sigma (t), P(t)) \text{ and} \]
\[ \dot{P}(t) = f (\sigma (t), P(t), \kappa (\sigma (t), P(t))) \quad \text{for } t \geq t_0, \]
\[ \frac{\partial D}{\partial t} (\sigma (t), P(t)) \]
\[ \frac{\partial D}{\partial x} D (\sigma (t), P(t)) \times f (\sigma (t), P(t), \kappa (\sigma (t), P(t))) \]
\[ + \frac{\partial D}{\partial t} (\sigma (t), P(t)), \]
\[ \text{for } t \geq t_0, \text{ the Lipschitzness of } \frac{\partial D}{\partial x}, \frac{\partial D}{\partial t}, \kappa \text{ and } f, \text{ and } \]}
\[ \text{ensure that the right hand side of } (68) \text{ and } (69) \text{ are Lipschitz and consequently } (P, \sigma) \in (C^1 (t_0, \infty) \times C^1 (t_0, \infty)). \]

VI. APPLICATION TO THE EXTRUSION PROCESS MODEL

From now, we recall the predictor feedback (47)–(46) for the compensation of the time- and state-dependent input delay in system (13) that we rewrite formally as
\[ \dot{x}(t) = f (t, x(t), U (t - D (t, x(t)))) \quad (71) \]
\[ f (t, x(t), U(t)) = -c(t) \Gamma (x(t), U(t)), \quad (72) \]
where \( D(t, x(t)) \) and \( c(t) \) are defined in (16) and (17), respectively, and
\[ \Gamma (x(t), U(t)) = \frac{\theta_2 x(t)}{(1 + \theta_2 x(t)) (1 - U(t))} - \frac{U(t)}{(1 - U(t))}, \quad (73) \]
The predictive feedback controller based on the piecewise exponential feedback law (22) is given by
\[ U(t) = \psi (P(t)), \quad (74) \]
\[ P(\theta) = x(t) + \int_{t-D(t,x(t))}^{\theta} \frac{f(s, P(s), U(s))}{1 - F(s, P(s), U(s))} \, ds \quad (75) \]
\[ \sigma(\theta) = t + \int_{t-D(t,x(t))}^{\theta} \frac{1}{1 - F(s, P(s), U(s))} \, ds, \quad (76) \]
for all \( t - D(t, x(t)) \leq \theta \leq t \). The function \( F \) defined in (49) for the system (71)–(72) is computed with the help of (16)–(17) as
\[ F(\sigma(t), P(t), U(t)) = \frac{\theta_1 \epsilon \omega \sin (\omega t)}{c^2(t)} (L - P(t)) \]
\[ + \Gamma (P(t), U(t)), \quad (77) \]
where \( \Gamma (P(t), U(t)) \) is defined in (73).

The parameters \( a_1(x^*) \) and \( a_2(x^*) \) of the feedback control law (74) are the solutions of (23) and (24) for an assigned slope function value at the set point that satisfies (31). \( P(t) = x (t + D(\sigma(t), P(t))) \) is the prediction of the state at the time when the current control will have an effect on the state. Recall that the implicit integral relation (75) is derived from the inversion of the time variable \( t \rightarrow t - D(t, x(t)) \) in \( t \rightarrow t + D(\sigma(t), P(t)) \) with the prediction time defined as \( \sigma(t) = t + D(\sigma(t), P(t)) \). The key point of the design is the feasibility condition \( F_\epsilon \) defined in (56), which ensures that the control action can reach the plant, namely, the delay rate is bounded by unity. The a priori satisfaction of (56) depends on the magnitude \( \epsilon \) and the angular frequency \( \omega \) of the periodic instability, and on the design parameters of the extruder.

Theorem 3. For any setpoint \( x^* \in (0, L) \) and any chosen setpoint slope \( S(x^*) \in \mathbb{R} \) satisfying \( S(x^*) \geq S_{\min}(x^*) \), where \( S_{\min}(x^*) \) is given by (37) and any initial condition \( x_0 \in (0, L) \),\[ \{U_0(\theta)| U_0(\theta) \in [0, 1], \text{ for all } \theta \in [-D(t_0, x_0)], \}
\]
\[ \text{taking the control gains } a_1(x^*) \text{ and } a_2(x^*) \text{ as solutions of (32) and (33), respectively, the closed-loop system consisting of the plant (71)–(73) with state } x(t), \text{ together with the control law (24)–(27)} \]
\[ \text{with actuator state } U(t + \theta), \theta \in [-D(t, x(t)), 0), \text{ is GAS at } x = x^*, U = v(x^*) \text{ if the parameters of the extruder model and the perturbation satisfy,} \]
\[ 0 \leq \frac{\epsilon \omega}{(1 - \epsilon)^2} < \frac{\theta_1 \theta_2}{(1 + \theta_2 L)^2}, \quad (79) \]
or,
\[ \frac{\theta_1 \theta_2}{(1 + \theta_2 L)^2} < \frac{\epsilon \omega}{(1 - \epsilon)^2} < \frac{\theta_1}{L}, \quad \text{and } \theta_2 < \frac{1}{L}, \quad (80) \]
or,
\[ \frac{\theta_1 \theta_2}{(1 + \theta_2 L)^2} < \frac{\epsilon \omega}{(1 - \epsilon)^2} < \frac{4 \theta_1 \theta_2}{(1 + \theta_2 L)^2}, \quad \text{and } \theta_2 > \frac{1}{L}, \quad (81) \]
where, \( \theta_1 \) and \( \theta_2 \) are defined in (44) and (15), respectively.

Proof: The proof of Theorem 2 is based on the Lyapunov-like condition (56) that must be satisfied a priori to guarantee the GAS property for any given \( x^* \in (0, L) \). In the following, we compute the function (49) for the time- and state-dependent input delay model of the extruder (71)–(73), (16), (17) in order to establish that the feasibility region as it is defined by (56) is the entire physical domain, namely, \( x \in (0, L) \) and \( U \in [0, 1] \). It holds that
\[ \frac{\partial D}{\partial t}(t, x) = \frac{\theta_1 \epsilon \omega \sin (\omega t)}{c^2(t)} (L - x), \quad (82) \]
\[ \frac{\partial D}{\partial x}(t, x) f(t, x, U(t)) = \Gamma (x, U), \quad (83) \]
where \( \Gamma (x, U) \) is given by (73). Since, \( 0 < \epsilon < 1 \), for all \( x \in (0, L) \) we get that \( \min \epsilon(t) = \theta_1 (1 - \epsilon) \), and hence,
\[ \frac{\partial D}{\partial t}(t, x) \leq \frac{\epsilon \omega (L - x)}{\theta_1 (1 - \epsilon)^2}. \quad (84) \]
The gradient of (73) with respect to the input \( U \) satisfies
\[ \nabla _U \Gamma (x, U) = -\frac{\theta_2}{(1 + \theta_2 x) (1 - U)^2}. \quad (85) \]
It follows that (73) is a strictly decreasing function of \( U \) which belongs to \([0, v_{\max}]\), for all \( x(t) \in [0, \infty) \) and
\[ \sup _{U \in [0, v_{\max}]} \Gamma (x, U) = \frac{\theta_2 x}{1 + \theta_2 x}. \quad (86) \]
The delay rate $\frac{dD}{dt}$ is uniformly bounded by unity, namely, the feasibility condition (56) is satisfied if and only if
$$\frac{\partial D}{\partial t}(t, x) + \frac{\partial D}{\partial x}(t, x)f(t, x, U) < 1. \tag{87}$$
for all $t \geq 0$, $x \in (0, L)$ and $U \in [0, 1)$. By (84) and (86), it follows that (87) is satisfied for all $x \in (0, L)$ and $U \in [0, v_{\text{max}}]$
1) if
$$\Lambda(x) = \frac{\epsilon \omega(L - x)}{\theta_1(1 - \epsilon)^2} + \frac{\theta_2 x}{1 + \theta_2 x}, \tag{88}$$
is a strictly increasing function since then its maximum over the domain $(0, L)$ satisfies
$$\Lambda(L) = \frac{\theta_2 L}{(1 + \theta_2 L^2)} < 1. \tag{89}$$
The derivative of the function (88) is written as
$$\Lambda'(x) = \frac{\theta_2}{1 + \theta_2 x^2} - \frac{\epsilon \omega}{\theta_1(1 - \epsilon)^2}, \tag{90}$$
and hence, since $x \in (0, L)$, (87) is guaranted if (79) holds.
2) if (88) is a decreasing function we deduce from (90) that
$$\frac{\epsilon \omega}{(1 - \epsilon)^2} > \theta_1 \theta_2, \tag{91}$$
and
$$\sup_{x \in [0, L]} \Lambda(x) = \frac{\epsilon \omega L}{\theta_1(1 - \epsilon)^2}. \tag{92}$$
Finally, the feasibility condition is satisfied if
$$\theta_1 \theta_2 < \frac{\epsilon \omega}{(1 - \epsilon)^2} < \frac{\theta_1}{L}. \tag{93}$$
One should notice that (93) necessarily restricts $\theta_2$ to satisfy
$$\theta_2 < \frac{1}{L}. \tag{94}$$
3) if (88) is an increasing function of $x$ on the interval $[0, x_1]$ and a decreasing function on $[x_1, L]$ such that $\Lambda(x_1) < 1$. The maximum value of $\Lambda$ is attained at $x_1$ satisfying
$$x_1 = (1 - \epsilon) \sqrt{\frac{\theta_1}{\theta_2 \epsilon \omega} - \frac{1}{\theta_2}}. \tag{95}$$
It becomes clear that (88) admits a unique maximum satisfying the feasibility condition if
$$0 < (1 - \epsilon) \sqrt{\frac{\theta_1}{\theta_2 \epsilon \omega} - \frac{1}{\theta_2}} < L, \tag{96}$$
and
$$\Lambda(x_1) < 1. \tag{97}$$
The relation (96) is equivalent to
$$\frac{\theta_1 \theta_2}{(1 + \theta_2 L^2)^2} < \frac{\epsilon \omega}{(1 - \epsilon)^2} < \theta_1 \theta_2. \tag{98}$$
Using (88), the inequality (97) leads to the following relation
$$\frac{\epsilon \omega}{(1 - \epsilon)^2} < \frac{4 \theta_1 \theta_2}{(1 + \theta_2 L)^2}. \tag{99}$$
For satisfying (98) and (99) we need to either impose (81) or the condition (95) with $\theta_2 < \frac{1}{L}$ which can be combined with (93) and (94) in order to derive (80).

**Remark 1.** For given values for the parameters of the extruder, namely, $\theta_1$, $\theta_2$, and $L$, the condition (79) is satisfied if $\epsilon$ or $\omega$ are sufficiently small. An increase of the magnitude of $\epsilon$ causes a decrease of the allowed $\omega$ and vice versa, as it is evident from (79)–(81). For given $\theta_1$, and $L$, the maximum bound of the perturbation parameters, namely, $\epsilon$ or $\omega$ is expressed in (79) as
$$\sup_{\theta_2 \in \mathbb{R}} \left\{ \frac{\theta_1 \theta_2}{(1 + \theta_2 L^2)^2} \right\} = \frac{\theta_1}{4L}, \quad \theta_2 = \frac{1}{L}. \tag{100}$$
Larger variations of $\epsilon$ and $\omega$ are possible, especially in the case in which $\theta_2$ is small, as it is evident from (80) and (81). However, for very large $\theta_2$, one can conclude from (79) and (81) that the allowable size of $\epsilon$ and $\omega$ is restricted. Moreover, from (79)–(81) one can conclude that the size of the allowable fluctuations of the transport speed in $\epsilon$ and $\omega$ is proportional to $\theta_1$ and inversely proportional to the extruder length $L$.

In physical terms, conditions (79)–(81) are mainly a correlation between the pressure and the “rotation” flow, namely, $\theta_2$ defined in (15) and $\theta_1$ defined in (14), respectively. We recall the expression of the net flow rate defined in (8) which is an increasing function of $\theta_2$ as it is shown in Fig. 4 Therefore, changes in $\theta_2$, by manipulating $K_d$, $B$, or $\rho_0$, the nozzle conductance, the screw resistance, and the melt density, respectively, affect the output flow rate $F_d$. For example, an increase in $\theta_2$ by increasing the nozzle conductance $K_d$, leads to an increase in the outflow rate. Note that $K_d$, which defines the nozzle opening, is directly related to the printing resolution, namely, the accuracy of the printing process. A large nozzle opening leads to an extrusion of a filament with a large diameter and consequently deteriorates the printer precision. Moreover, from (80) and (15), it can be also seen that the “robustness” of the controller depends on the material thickness, namely, the mass density $\rho$: a thicker material is less sensitive to large fluctuations of the transport speed under the predictor feedback control law. The parameter $B$ in the expression of $\theta_2$ in (15) is given by
$$B = \frac{WH^3}{12}, \tag{101}$$
where $H$ is the approximate depth of screw channel from the screw thread root to the barrel internal surface and $W$ is the width of screw channel. Consequently, changes in $\theta_2$ due to the changes in $B$ affect also the parameter $\theta_1$, since the screw pitch value $\xi$ also depends directly on $W$.

Relations (79)–(81) show that an increase in $\theta_1$, namely, an increase of the material convection speed, by enabling a large screw pitch $\xi$ or a high screw speed $N_0$, improves the “robustness” of the controller in some way and allows for a system that supports broader changes of the convection velocity in frequency and amplitude. Note that a sharp
increase in the rotational screw speed $N_0$ results in material overload and clogging problems and has a major effect on the residence time that is the critical time during which the material should be heated to have good properties before being evicted through the nozzle. Particularly, the extruded filament homogeneity is directly related to the residence time and to the process of solidification after layers deposition in 3D printers. In addition, an increase in $\theta_1$ in the screw speed $N_0$, increases the thermal energy in the extruder chamber due to the material shearing and decreases the viscosity of the melt. In that case, a rapid feeding of the extruder with granular material by applying a more aggressive “Bang-Bang” control action absorbs the excess heat in the system. Maintaining a reasonable temperature inside the barrel is essential because an excessive overheating of the system burns the polymer or produces poor extrusion. Generally, the conventional extrusion processes are equipped with a cooling system to compensate for the heat generated by the mechanical shearing effect that is proportional to the screw speed.

In general, the nozzle and the screw designs are directly related to the predictor feedback control design and for achieving high performances for the closed-loop system the scale of the extruder should be neatly chosen. For instance, the aggressiveness of the controller is influenced by the choice of $\theta_2$ since the minimum value of the slope at the set point $S_{\text{min}}(x^*)$ defined in (31) depends on this parameter. Moreover, the entire process operates with an extruder head that moves very fast to print filament lines layer upon layer on a moving platform. A sufficiently light extruder head with small nozzle opening $K_d$ and a small length $L$ that operates at a sufficiently high screw speed $N_0$ is needed to ensure a high rate of extrusion with a high precision.

$L = 0.2$ m and the system is supposed to settle at $x_0 = 0.1$m at the initial time. The value of the slope function is set to $S(x^*) = S_{\text{min}}(x^*) + 30$. The simulations show the dynamics of the input filling ratio $U(t)$, the interface position $x(t)$, the predictor state $P(t)$ and the delay function $D(t, x(t))$. Different cases including the open-loop dynamics, both uncompensated and compensated delay control laws are simulated for $\{\epsilon = 0.1, \omega = 3.5 \text{ rad/s}\}$ and $\{\epsilon = 0.4, \omega = 0.4 \text{ rad/s}\}$. It is clear that, the uncompensated input leads to a limit cycle and the compensated closed-loop control allows faster convergence than the open loop control. Also, as it is shown in Figure 9 the feasibility condition is satisfied in both presented simulation results.

Fig. 4. The nozzle flow rate is an increasing function of $\theta_2$.

VII. SIMULATIONS

A. Time- and state-dependent input delay compensation

The setpoint is chosen as $x^* = 0.16$ m which corresponds to a desired nozzle output flow rate as indicated by the equation (8). The initial position of the moving interface is set to $x_0 = 0.1$ m, the total length of the extruder is

B. The state-dependent input delay compensator for the model with constant viscosity

The following simulation results show the stabilization of the model described by (11) with the state-dependent input delay predictor feedback law (20). Defining (11) as

$$
\dot{x}(t) = f(x, U(t - D_s(x(t))))
$$

and

$$
D_s(x(t)) = \frac{L - x(t)}{\theta_1},
$$

where

$$
f(x(t), U(t)) = -\theta_1 \Gamma(x(t), U(t)),
$$

(103)

(104)

(102)

Fig. 5. Compensation of the time- and state-dependent input delay-(a).

Fig. 6. Compensation of the time- and state-dependent input delay-(b).
\[ \theta_1 \text{ is the nominal transport velocity of the material defined in } [14], \text{ and the function } \Gamma(x(t), U(t)) \text{ is given by } [13]. \]

The predictor feedback controller is written as

\[
U(t) = v(P_x(t)),
\]

\[
P_x(t) = x(t) + \int_{t-D_n(x(t))}^{t} \frac{f(P_x(\mu), U(\mu))}{1 - F_x(P_x(\mu), U(\mu))} d\mu.
\]

where for all \( t - D_n(x(t)) \leq \mu \leq t \)

\[
F_x(P_x(\mu), U(\mu)) = \frac{\partial D_n}{\partial x}(P_x(\mu))f(P_x(\mu), U(\mu)).
\]

By specializing Theorem 2 to the case \( \epsilon = 0 \) it can be shown that the predictor feedback law \([105]-[107]\) renders system \([102]-[104]\) GAS (in the physical domain) at any given equilibrium \( x^\ast \).

More precisely, both the predictor state, \( \hat{P}(t) \) and the delay function, \( \hat{D}(x(t)) \), are estimates of the actual prediction state, namely, \( P(t) \) and delay function, namely, \( D(t, x(t)) \), that are described by \([75]-[16], [17]\), respectively. For implementing the controller \([108]-[109]\), the “actual” feasibility condition, defined in \([79]-[80], [81]\) have to hold, in order to guarantee that the predictor actually “kicks in”. In addition, we assume that the following condition, which guarantees that the denominator in \([109]\) remains always positive (and hence, the controller remains bounded) is satisfied

\[
\frac{\theta_2 P(\theta)}{(1 + \theta_2 P(\theta))(1 - U(\theta))} - \frac{U(\theta)}{(1 - U(\theta))} < 1,
\]

\[ \theta \in [t - \hat{D}(x(t)), t]. \]

Note that under strictly positive physical parameters \( B, p_0, \) and \( K_d \), and for \( U \in [0, v_{\text{max}}] \), relation \([113]\) is satisfied whenever \( \hat{P}(\theta) \in [0, \infty] \), for all \( \theta \in [t - \hat{D}(x(t)), t] \).

The simulation results in Fig. 8 illustrate that the state-dependent input delay compensator can handle small time-varying uncertainties on the vector field \([112]\) and the delay function \([111]\), as described by \([72]-[73] \) and \([16]-[17]\), respectively. An increase in the necessary control effort to drive the system to the setpoint is also denoted and the rate of convergence decreases compared to the time- and state-dependent predictor feedback \([74]-[76]\) shown in Fig. 6.

C. Control with a state-dependent input delay compensator

We deal with the case in which the time variations of the transport speed are unknown and consider the closed-loop system consisting of the plant \([13]\) with an actual delay \( D(t, x(t)) \), given in \([16], [17]\), together with a state-dependent input delay predictor feedback defined by

\[
U(t) = v(\hat{P}(t)),
\]

\[
\hat{P}(t) = x(t) + \int_{t-\hat{D}(x(t))}^{t} \hat{f}(\hat{P}(s), U(s)) ds,
\]

where,

\[
\hat{f}(\hat{P}(s), U(s)) = \frac{\partial \hat{D}}{\partial x}(\hat{P}(s))\hat{f}(\hat{P}(s), U(s)).
\]

With an estimated delay function defined as

\[
\hat{D}(x(t)) = \frac{L - x(t)}{\xi N_0},
\]

and the nominal vector field,

\[
\hat{f}(x(t), U(t)) = \theta_1 \left[ \frac{\theta_2 x(t)}{(1 + \theta_2 x(t))(1 - U(t))} + \frac{U(t)}{1 - U(t)} \right], \quad U(t) \in [0, 1].
\]
VIII. Conclusions

This paper is devoted to the stabilization of a screw-extrusion process. A coupled PDE-ODE model is used to derive a state-dependent input delay system describing the melt convection in the extruder chamber for an isothermal case. The extension of the aforementioned model to the non-isothermal case is proposed introducing a periodic time-dependent function in the state-dependent input delay function. Next, we design a predictor feedback controller to compensate the state- and time-dependent input delay and establish the GAS of any setpoint with respect to the physical domain under physical and design restrictions. The delay compensator is constructed with a nominal Bang-Bang-like controller that ensures the GES of the delay-free plant.

It is clear that the delay function model for the non-isothermal extrusion process should depend on the rheological properties of the extruded polymer. In general, a fairly accurate estimation of the material friction and the viscosity behavior is extremely hard to achieve in such processes due to the change in material composition, and the strong interaction between the heat and mass transfer phenomena. An interesting future work would be to consider an unknown time-dependent perturbation acting on the polymer convection speed. As it is shown in the simulation results, the state-time-dependent perturbation acting on the polymer convection speed. As it is shown in the simulation results, the state-time-dependent perturbation acting on the polymer convection speed.

Next, we design a predictor feedback controller to compensate the state- and time-dependent input delay and establish the GAS of the delay-free plant.

A coupled PDE-ODE model is used to derive a state-dependent input delay system (43) with the controller (47)–(46) into the following target system

\[ \dot{x}(t) = f(t, x(t), \kappa(t, x(t))) + W(t - D(t, x(t))) \]  
\[ W(t) = 0. \]  

Proof: The proof of Lemma [1] is based on a direct verification considering \( P(t - D(t, x(t))) = x(t) \) and \( \sigma(t - D(t, x(t))) = t \) in the original system (43).

Lemma 2. (Inverse Backstepping Transformation) The inverse of the infinite dimensional backstepping transformation (A-I) is defined for all \( t - D(t, x(t)) \leq \theta \leq t \) by

\[ U(\theta) = W(\theta) + \kappa(\sigma(\theta), \Pi(\theta)), \]  

with

\[ \Pi(\theta) = \int_{\phi(t)}^{\theta} \frac{f(\sigma(s), \Pi(s), \kappa(\sigma(s), \Pi(s))) + W(s)}{1 - F(\sigma(s), \Pi(s), W(s))} ds + x(t) \]  
\[ F(\sigma(s), \Pi(s), W(s)) = \frac{\partial D}{\partial t}(\sigma(s), \Pi(s)) + \frac{\partial D}{\partial x}(\sigma(s), \Pi(s)) \times f(\sigma(s), \Pi(s), \kappa(\sigma(s), \Pi(s))) + W(s), \]  
\[ \sigma(\theta) = t + \int_{\phi(t)}^{\theta} \frac{1}{1 - F(\sigma(s), \Pi(s), W(s))} ds, \]  
\[ \phi(t) = t - D(t, x(t)). \]  

Proof: Direct verification considering that \( P(\theta) = \Pi(\theta) \) and \( \sigma(\theta) = \sigma(\theta) \) for all \( t - D(t, x(t)) \leq \theta \leq t \). We refer to \( P(\theta) \) as the plant-predictor system and \( \Pi(\theta) \) as the target-predictor system, respectively. However, they play different roles because they are driven by different inputs (\( U \) versus \( W \)).

Lemma 3. (Stability of the Target System) For any positive constant \( g \), there exist a class \( K_{\infty} \) function \( \rho_* \) and a class \( KL \) function \( \beta \) such that for all solutions of the system satisfying the feasibility condition (56), the following holds:

\[ \Xi(t) \leq \beta(\rho_*(\Xi(t_0)), t - t_0), \quad t \geq t_0 \]  
\[ \Xi(t) = |x(t)| + \sup_{t-D(t,x(t))\leq\theta\leq t} |W(\theta)| \]  

where

\[ \rho_*(s) = \frac{e^{\frac{t}{1-\epsilon}}(e^{\frac{t}{1-\epsilon}}(c_1 + \mu_4(s))_s)}{1-e^{\frac{t}{1-\epsilon}}} \]  

Proof: Based on the input-to-state stability of \( \dot{x} = f(t, x, \kappa(t, x) + \omega) \) with respect to \( \omega \), namely, Assumption [2], there exist a smooth function \( S(t, x(t)) : [t_0, \infty) \times \mathbb{R}^n \rightarrow \mathbb{R}_+ \) and class \( K_{\infty} \) functions \( \alpha_1, \alpha_2, \alpha_3, \alpha_4 \) such that

\[ \alpha_1(|x(t)|) \leq S(t, x(t)) \leq \alpha_2(|x(t)|), \]  
\[ \dot{S}(t, x(t)) \leq -\alpha_3(|x(t)|) + \alpha_4(|W(t - D(t, x(t)))|), \]  

with

\[ \dot{S}(t, x(t)) = \frac{\partial S(t, x(t))}{\partial t} + \frac{\partial S(t, x(t))}{\partial x} \times f(t, x(t), \kappa(t, x(t)) + W(t - D(t, x(t)))) \]  

APPENDIX

We recall Lemmas 1–8 from [20], which are applied to the nonlinear time- and state-dependent input delay system as Lemmas [21]–[25] for the proof of Theorem [2].

Lemma 1. (Backstepping Transformation of the Actuator State) The infinite dimensional backstepping transform of the actuator state given by

\[ W(\theta) = U(\theta) - \kappa(\sigma(\theta), P(\theta)), \]  

for all \( t - D(t, x(t)) \leq \theta \leq t \), allows to transform the system (43) with the controller (47)–(46) into the following target system:

\[ \dot{x}(t) = f(t, x(t), \kappa(t, x(t))) + W(t - D(t, x(t))) \]  
\[ W(t) = 0. \]  

Proof: The proof of Lemma [1] is based on a direct verification considering \( P(t - D(t, x(t))) = x(t) \) and \( \sigma(t - D(t, x(t))) = t \) in the original system (43).
Let us define the Lyapunov function for the target system (A-3) and (A-3) as
\[ V(t) = S(t, x(t)) + k \int_0^{L(t)} \frac{\alpha(r)}{r} dr, \]  
(A-16)
where
\[ L(t) = \sup_{t-D(t,x(t)) \leq \theta \leq t} \left| e^{g(1+\sigma(\theta) - t)} W(\theta) \right| \]
\[ = \lim_{n \to \infty} \left( \int_{t-D(t,x(t))}^{t} e^{2ng(1+\sigma(\theta) - t)} W(\theta)^2 d\theta \right)^{1/2n}, \]
(A-17)
with \( g > 0 \). Let us upperbound and lowerbound (A-17) in terms of:
\[ \sup_{t-D(t,x(t)) \leq \theta \leq t} |W(\theta)| \]  
(A-18)

- **Upperbound of \( L(t) \):** Using the feasibility condition (56) and (55), we deduce
\[ \dot{\sigma}(\theta) \leq -\frac{1}{1-c}. \]  
(A-19)
By integration of (A-19) on \([t-D(t,x(t)) \theta] with \( \sigma(t-D(t,x(t))) = t \), we derive the inequality
\[ 1 + \sigma(\theta) - t \leq \frac{1}{1-c} (1 + D(t,x(t))), \]
\[ \forall \ t - D(t,x(t)) \leq \theta \leq t. \]  
(A-20)
From Assumption 3 and (60), the following inequality holds:
\[ L(t) \leq e^{2g \left(1+\varepsilon_1+\mu_4(|x|)\right)} \sup_{t-D(t,x(t)) \leq \theta \leq t} |W(\theta)|. \]  
(A-21)

- **Lowerbound of \( L(t) \):** Similarly, using the fact that \( \sigma(t-D(t,x(t))) = t \), with \( \theta(t) \) being an increasing function, we obtain
\[ 1 + \sigma(\theta) - t \geq 1, \]  
(A-22)
and hence
\[ L(t) \geq e^{g} \sup_{t-D(t,x(t)) \leq \theta \leq t} |W(\theta)| \]
\[ \forall \ t - D(t,x(t)) \leq \theta \leq t. \]  
(A-23)
The time derivative of (A-17) is
\[ \dot{L}(t) = \lim_{n \to \infty} \frac{1}{2n} \left( \int_{\phi(t)}^{t} e^{2ng(1+\sigma(\theta) - t)} W(\theta)^2 d\theta \right)^{1/2n-1} \]
\[ \times \left\{ - \frac{dD(t,x(t))}{dt} e^{2ng} W(t-D(t,x(t)))^{2n} \right\} \]
\[ - 2ng \int_{\phi(t)}^{t} e^{2ng(1+\sigma(\theta) - t)} W(\theta)^2 d\theta \}, \]  
(A-24)
where \( \phi(t) = t - D(t,x(t)) \). By (56), it is clear that
\[ \frac{dD(t,x(t))}{dt} < 1 \]  
and hence
\[ \dot{L}(t) \leq -gL(t). \]  
(A-25)
Computing the derivative of the Lyapunov function (A-16) as
\[ \dot{V}(t) = \dot{S}(t,x(t)) + k\dot{L}(t) \frac{\alpha_4(L(t))}{L(t)}, \]  
(A-26)
we deduce
\[ \dot{V}(t) \leq \dot{S}(t,x(t)) - k\alpha_4(L(t)). \]  
(A-27)
Using the boundness of \( S(x(t)) \), (A-14), the following inequality holds:
\[ \dot{V}(t) \leq -\alpha_3(|x(t)|) + \alpha_4(|W(t-D(t,x(t)))|) - k\alpha_4(L(t)). \]  
(A-28)
Imposing \( k = g^{-1} \), by (A-23), we derive the inequality
\[ \dot{V}(t) \leq -\alpha_3(|x(t)|) - \alpha_4(L(t)), \]  
(A-29)
and with (A-13), (A-16) and (A-17), we conclude that there exists a \( K \) function \( \gamma_1 \) such that
\[ \dot{V}(t) \leq -\gamma_1(V(t)). \]  
(A-30)
By the comparison principle, there exists a class \( KL \) function \( \beta \) such that
\[ V(t) \leq \beta(V(t_0), t - t_0). \]  
(A-31)
From (A-13) and (A-16) and the properties of class \( KL \) functions, we finally get
\[ |P(\theta)| \leq \rho \left( |x(t)| + \sup_{t-D(t,x(t)) \leq s \leq t} |U(s)| \right). \]  
(A-32)

**Lemma 4. (Bound of the Predictor in Terms of Actuator State)** There exists a class \( K_\infty \) function \( \rho \) such that for all the solutions of the system satisfying the feasibility condition (56), the following holds for all \( t - D(t,x(t)) \leq \theta \leq t \)
\[ |P(\theta)| \leq \rho \left( |x(t)| + \sup_{t-D(t,x(t)) \leq s \leq t} |U(s)| \right). \]  
(A-33)

**Proof:** Differentiating (47), we deduce the following relation for all \( t - D(t,x(t)) \leq \theta \leq t \)
\[ \frac{dP(\theta)}{d\theta} = \frac{f(\sigma(\theta), P(\theta), U(\theta))}{1 - F(\sigma(\theta), P(\theta), U(\theta))}, \]
\[ F(\sigma(\theta), P(\theta), U(\theta)) = \frac{\partial D}{\partial \sigma} \frac{\partial D}{\partial P} \frac{\partial}{\partial U} f(\sigma(\theta), P(\theta), U(\theta)) + \frac{\partial D}{\partial \sigma} (\sigma(\theta), P(\theta)), \]  
(A-34)
and with the change of variable \( y = \sigma(\theta), \) (34) may be rewritten as:
\[ \frac{dP(\phi(y))}{dy} = f (y, P(\phi(y)), U(y-D(\phi(y))), \]  
\[ t \leq y \leq \sigma(t). \]  
(A-35)
From Assumption 1, we get that
\[ \frac{dR(x,y, P(\phi(y)))}{d\theta} \leq \hat{\sigma}(\theta) (R(x, P(\phi(y)))) \]
\[ + \mu_3 (|U(y-D(\phi(y))))\), \]  
(A-36)
for all \( t \leq y \leq \sigma(t) \) and using the feasibility condition (56), we deduce, for all \( t - D(t,x(t)) \leq \theta \leq t \)
\[ \frac{dR(\sigma(\theta), P(\theta))}{d\theta} \leq \frac{1}{1-c} (R(\sigma(\theta), P(\theta)) + \mu_3(|U(\theta)|)). \]  
(A-37)
By Assumption 3 and the comparison principle, we obtain
\[
R(\sigma(\theta), P(\theta)) \leq e^{\frac{1}{\sigma_{\text{KL}}}(c_{1}+\mu_{4}(|x|))} \left( R(t, x(t)) + \sup_{t-D(t, x(t))\leq s\leq t} \mu_{3}(|U(s)|) \right),
\]

\[
t - D(t, x(t)) \leq \theta \leq t. \tag{A-38}
\]

With the standard properties of class $\mathcal{KL}$ functions the Lemma 4 is deduced and the class $\mathcal{KL}$ function $\rho$ is written as:
\[
\rho(s) = \mu_{1}^{-1} \left( (\mu_{2}(s) + \mu_{3}(s)) e^{\frac{1}{\sigma_{\text{KL}}}(c_{1}+\mu_{4}(s))} \right). \tag{A-39}
\]

Lemma 5. (Bound of the Predictor in Terms of Transformed Actuator State) There exists a class $\mathcal{K}$ function $\psi$ such that for all the solutions of the system satisfying the feasibility condition (56), the following holds:
\[
|\Pi(\theta)| \leq \psi \left( |x(t)| + \sup_{t-D(t, x(t))\leq s\leq t} |W(s)| \right), \tag{A-40}
\]
for all $t-D(t, x(t)) \leq \theta \leq t$.

**Proof:** The plant $\dot{X}(t) = f(t, x(t), \kappa(t, x(t)) + \omega(t))$ satisfying the uniform input-to-state stability property with respect to $\omega$, and the function $\kappa$ being locally Lipschitz in both arguments and uniformly bounded with respect to its first argument, there exist a class $\mathcal{KL}$ function $\beta_{2}$ and a class $\mathcal{K}$ function $\psi_{1}$ such that for all $\tau \geq t_{0}$
\[
Y(\tau) \leq \beta_{2}(|Y(t_{0})|, \tau - t_{0}) + \psi_{1} \left( \sup_{s \geq t_{0}} |\omega(s)| \right), \tag{A-41}
\]
with
\[
\dot{Y}(\tau) = f \left( Y(\tau), \kappa(\tau, Y(\tau)) + \omega(\tau) \right). \tag{A-42}
\]

Now, we consider the change of variable $y = \sigma(\theta)$ and write the predictor of the target system (A-5) as
\[
\frac{d\Pi(\phi(y))}{dy} = f \left( y, \Pi(\phi(y)), \kappa(y, \Pi(\phi(y))) + \omega(\phi(y)) \right),
\]
\[
t \leq y \leq \sigma(t). \tag{A-43}
\]

Using (A-42), we derive the following relation
\[
|\Pi(\theta)| \leq \psi_{2}(|x(t)|) + \psi_{1} \left( \sup_{t-D(t, x(t))\leq s\leq t} |W(s)| \right), \tag{A-44}
\]
for all $t-D(t, x(t)) \leq \theta \leq t$ with a class $\mathcal{K}$ function $\psi_{2}(s) = \beta_{2}(s, 0)$ . Using the properties of class $\mathcal{K}$ functions, (A-40) is deduced with $\psi(s) = \psi_{1}(s) + \psi_{2}(s)$.

Lemma 6. (Equivalence of the Norms of the Original and the target system)
There exist class $\mathcal{KL}$ functions $\rho_{1}$, $\mu_{7}$ such that for all the solutions of the system satisfying the feasibility condition (56), and for all $t \geq t_{0}$, the following hold:
\[
\Omega(t) \leq \mu_{7}^{-1}(\Xi(t)), \tag{A-45}
\]
\[
\Xi(t) \leq \rho_{1}(\Omega(t)), \tag{A-46}
\]
where $\Omega$ and $\Xi$ are defined in (64) and (A-11), respectively.

**Proof:** Using the inverse transformation (A-4) and the bound (A-40), we derive (A-45) with
\[
\mu_{7}^{-1}(s) = s + \hat{\rho}(\psi(s)), \tag{A-47}
\]
and from the direct transformation (A-1) together with the bound (A-33), we deduce (A-46), where $\rho_{1}$ is defined as
\[
\rho_{1}(s) = s + \hat{\rho}(\rho(s)). \tag{A-48}
\]

Lemma 7. (Ball Around the Origin Within the Feasibility Region) There exists a positive constant $\tilde{\gamma}$ such that for all the solutions of the system that satisfy
\[
|x(t)| + \sup_{t-D(t, x(t))\leq s\leq t} |U(\theta)| < \tilde{\gamma}, \tag{A-49}
\]
the feasibility condition (56) is satisfied.

**Proof:** From (45) we derive the following inequality
\[
|f(t, x(t), U(t - D(t, x(t))))| \leq \hat{\alpha} \left( |x(t)| + \sup_{t-D(t, x(t))\leq s\leq t} |U(s)| \right), \tag{A-50}
\]
Recalling the relations (61) and (62) of Assumption 3 we deduce that for all $\theta \in [t - D(t, x(t)), t]$ and $c \in [0, 1]$, if a solution satisfies
\[
c_{3} + \mu_{5}(|P(\theta)|) + (c_{2} + \mu_{6}(|P(\theta)|)) \hat{\alpha} \left( |P(\theta)| + \sup_{t-D(t, x(t))\leq s\leq t} |U(s)| \right) < c, \tag{A-51}
\]
then it also satisfies (56).

Using Lemma 4 we conclude that (A-51) is satisfied if the following holds
\[
(c_{2} + \mu_{6}(\rho(\Omega(t)))) \hat{\alpha} (\rho(\Omega(t)) + \Omega(t)) + \mu_{5}(\rho(\Omega(t))) < c - c_{3}. \tag{A-52}
\]
Let us define a class $\mathcal{KL}$ function $\rho_{c}$ as
\[
\rho_{c}(s) = \mu_{5}(\rho(s)) + (c_{3} + \mu_{6}(\rho(s))) \hat{\alpha} (\rho(s) + s). \tag{A-53}
\]

It follows that
\[
\tilde{\gamma} = \rho_{c}^{-1}(c - c_{3}). \tag{A-54}
\]

Lemma 8. (Estimate of the Region of Attraction) There exists a class $\mathcal{K}$ function $\psi_{\text{RDA}}$ such that for all initial conditions of the closed-loop system that satisfy relation (63), the solutions of the system satisfy (A-49) for $c \in [0, 1]$, and hence, satisfy (56).

**Proof:** Using Lemma 6 and (A-10), the following holds:
\[
\Omega(t) \leq \mu_{7}^{-1} \left( \beta(\rho_{c}(\rho_{1}(\Omega(t_{0}))), t - t_{0}) \right), \tag{A-55}
\]
where $\Omega$ is defined in (64). Introducing the class $\mathcal{KL}$ function $\mu_{9}(s) = \mu_{7}^{-1}(\beta(s, 0))$, we derive the inequality
\[
\Omega(t) \leq \mu_{9}(\rho_{c}(\rho_{1}(\Omega(t_{0})))) \tag{A-56}
\]

Hence, for all initial conditions that satisfy the bound (56) with any class $K$ choice

$$\psi_{RoA}(c - c_0) \leq \rho_1^{-1}(\rho_2^{-1}(\rho_3^{-1}(c - c_3))),$$  \hspace{1cm} (A-57)

the solutions satisfy (A-49). Moreover, for all of those initial conditions, the solutions verify (64), for all $\theta > t_0 - D(t_0, z(t_0))$.

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