PDE Control of Heat Exchangers by Input-Output Linearization Approach

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Abstract: Heat exchangers are widely used in industry for cooling or heating fluids. Their tight control is important for optimal energy management. The dynamical behavior of heat exchangers is described by a set of coupled partial differential equations describing the profiles of internal and external temperatures, which makes their control a challenging problem. The present chapter discusses the geometric control of both counter-current and co-current flow heat exchangers. The simplest control approach would lie on a priori discretization of the PDEs resulting in a set of ODEs that can be tackled by traditional control laws but that does not preserve the fundamental control properties of the heat exchanger (controllability, observability and stability). In the following, the PDEs are directly used to derive the control laws and the discretization is performed a posteriori only for simulation needs. The

design of control laws is based on the concept of characteristic index in the framework of input-output linearization and the closed loop stability is demonstrated using some tools from semigroup theory. The control objective consists in controlling the outlet cold fluid temperature by manipulating the inlet hot fluid temperature. Then, robustness with respect to parameter uncertainty and practical implementation are discussed and a general control strategy is introduced. Both tracking and disturbance rejection performances of the developed control laws are assessed by numerical simulation.

Keywords: Distributed parameter system, partial differential equation, heat exchanger, boundary control, geometric control, characteristic index, input-output linearization.

1 Introduction

Heating or cooling of liquid and gaseous fluids are two operations involved in many engineering applications. These operations are achieved using heat exchangers that transfer heat from a hot fluid to a cooler one (Al-Dawery et al., 2012; Kakaş and Liu, 2002). Heat exchangers can be classified according to the geometry construction (tubes and plates), to the transfer processes (direct and indirect contacts), to the flow arrangement (counter, parallel and cross-flows) and to the fluid phase (liquid-liquid and gas-liquid) (Kakaş and Liu, 2002). Tube heat exchangers are widely used in industry due to their ability to ensure a wide range of operating temperatures (Al-Dawery et al., 2012; Fratzczak et al., 2018).

In a heat exchanger, convection is usually the dominant form of heat transfer compared to diffusion, which is negligible (Arbaoui et al., 2007; Friedly, 1972; Ray and Ogunnaike, 1994). Thus, the dynamic behavior of a heat exchanger is often captured by two linear partial differential equations (PDEs) of hyperbolic type with one spatial variable. These equations, obtained by writing the energy balances separately for both fluids, are weakly coupled via the source terms (difference of temperatures between the two fluids) (Friedly, 1972; Ray and Ogunnaike, 1994). Each PDE describes the spatio-temporal evolution of the temperature of the related fluid. Thus, a heat exchanger is typically an example of a 2×2 (2 equations with 2 dependent variables, i.e. time and space) hyperbolic system issued from balance laws (Bartecki, 2016).

Heat exchangers are characterized by two flow rates and two inlet temperatures of fluids. These variables can be used as manipulated variables to control the fluid temperatures at the outlet (Maidi et al., 2009). Note that the fundamental control-theoretic properties (controllability, observability and

stability) depend on the choice of manipulated variables (Alotaibi et al., 2004). By manipulating the inlet temperatures, the heat exchanger is a linear distributed parameter system (DPS), consequently control-theoretic properties can be easily analyzed using the powerful semi-group theory (Grabowski, 2007; Kunitatsu, 1998; Sano, 2007). In the case of flow rate manipulation, the heat exchanger is a nonlinear (bilinear) DPS (Bühler and Franke, 1980) and both control design and control-theoretic properties analysis are difficult tasks.

Essentially, to deal with control design for heat exchangers, two main approaches (Christofides, 2001; Ray, 1989), indirect or direct, are possible. The indirect approach (early lumping) consists in reducing the PDEs to a set of ordinary differential equations (ODEs) by approximating either the PDEs themselves (by discretization) or their solutions (Li and Qi, 2010). This approach allows to exploit various well established linear control theories of lumped parameter systems (LPSs). Nevertheless, the indirect approach masks the distributed nature of the heat exchanger and often the fundamental control-theoretic properties are lost due to the approximation process (Christofides, 2001; Ray, 1989; Singh and Hahn, 2007). In addition, this approach leads to a high dimensional controller with limited performance and difficult practical implementation issues (Christofides and Daoutidis, 1996). The direct approach (late lumping) consists in using directly the PDEs model to design the controller without any previous approximation. The direct approach preserves control-theoretic properties of the heat exchanger and leads to an infinite dimensional controller that is of distributed nature, which enhances the control quality (Christofides and Daoutidis, 1996; Maidi et al., 2009, 2010). Nevertheless, the practical implementation needs the use of an observer to reconstruct the whole state from the accessible measurements (Maidi et al., 2009, 2010).

Heat exchanger control has been extensively investigated in the literature and many control design techniques are proposed. The heat exchanger control literature reveals that most contributions deal with the control of heat exchanger by the indirect approach whereas the direct approach is less addressed. This can be explained by the fact that designing the controller by using directly the PDEs model is a difficult task.

In the following, different control methods of heat exchangers using both indirect and direct approaches are briefly reviewed. First, consider control methods based on the approximate or reduced model. Malleswararao and Chidambaram (1992) proposed a model reference non-linear controller with PID actions. Fischer et al. (1998) and Mazinan and Sadati (2010) developed fuzzy predictive controllers.

Fuzzy control and neural network approaches have been adopted by Al-Dawery et al. (2012) and Bitanti and Piroddi (1997); Vasičkaninová et al. (2011), respectively. Based on the convexity property, Abdelghani-Idrissi et al. (2001) developed a predictive functional controller. Using an approximate first order model, an adaptive version of a predictive functional controller is implemented by Arbaoui et al. (2007). A control strategy combining neural network predictive and fuzzy controllers is proposed by Vasičkaninová and Bakošová (2015). A sliding mode controller is developed in Almutairi and Zribi (2012). Other advanced robust control methods has also proposed by Dulău et al. (2015); Gauthier and Xu (1991); Kanoh et al. (1997); Vasičkaninová and Bakošová (2016) and Vasičkaninová et al. (2018).

Considering the direct approach, few contributions have been reported in the literature. Hanczyc and Palazoglu (1995) designed a sliding mode control based on the equivalent LPS model obtained using the method of characteristics. Robust control theory has been generalized to DPSs by Pohjolainen and Lätti (1983), thanks to the operators theory, and a flow adaptive robust controller is developed. Using semigroup theory, Sano (2015) proposed a residual mode filter-based controller to enhance the stability under a boundary actuation and boundary sensing. In Aulisa et al. (2016), the nonlinear tracking problem is solved using velocity control configuration. Fratzak et al. (2018) developed a distributed parameter balance-based adaptive controller and performed a practical validation on a laboratory heat exchanger. Boundary control based on the input-output linearization approach has been solved by Maidi et al. (2009, 2010).

Geometric control is an interesting late lumping approach that is well adapted for control problem of DPSs (Christofides and Daoutidis, 1996; Maidi and Corriou, 2011b, 2014a,b; Shang et al., 2005; Wu and Liou, 2001). The present chapter is devoted to the direct control design of geometric control laws for counter-current and co-current heat exchangers. The chapter starts with the control problem formulation. Then, the design approaches based on the PDEs model are presented. The stability and practical implementation issues of the developed control laws are also discussed. Finally, the performance of the presented control laws are evaluated via numerical simulation. The chapter ends with a conclusion.

2 Heat exchanger control problem

The considered heat exchanger is a double pipe used to heat the cold fluid circulating in the internal tube by means of the hot fluid circulating in the external tube as depicted in Figure 1. In the counter-

current heat exchanger, the hot and cold fluids flow in opposite directions (Fig. 1a) while in the case of a co-current heat exchanger (Fig. 1b), both fluids flow in the same direction. Under some reasonable assumptions (Arbaoui et al., 2007; Friedly, 1972; Ray and Ogunnaike, 1994), the dynamical behavior of this double pipe heat exchanger is described by the following 2×2 hyperbolic of balance laws (Friedly, 1972):

$$\frac{\partial T_h(z, t)}{\partial t} = -\vartheta_h \frac{\partial T_h(z, t)}{\partial z} + \alpha_h (T_c(z, t) - T_h(z, t)) \quad (1)$$

$$\frac{\partial T_c(z, t)}{\partial t} = \beta \vartheta_c \frac{\partial T_c(z, t)}{\partial z} + \alpha_c (T_h(z, t) - T_c(z, t)) \quad (2)$$

$$T_h \left(\frac{(1+\beta)}{2} l, t \right) = \frac{(1-\beta)}{2} T_{h_0}(t) + \frac{(1+\beta)}{2} T_{h_l}(t) \quad (3)$$

$$T_c(0, t) = T_{c_0}(t) \quad (4)$$

$$T_h(z, 0) = T_h^*(z) \quad (5)$$

$$T_c(z, 0) = T_c^*(z) \quad (6)$$

In the heat exchanger model (1)–(6), $t \in \mathbb{R}^+$ and $z \in \Omega$ are the time and space independent variables, respectively. l is the length of the heat exchanger and $\Omega = [0, l]$ is the space domain. T_h and T_c denote the temperatures of the hot and cold fluids, respectively, and T_h^* and T_c^* are their initial temperature profiles. T_{c_0} is the inlet temperature of the cold fluid. T_{h_0} and T_{h_l} denote the inlet temperatures of the hot fluid for the co-current and counter-current heat exchangers, respectively. α_h and α_c are the heat transfer coefficients, ϑ_h and ϑ_c are the flow rates of the fluids. β is a constant parameter that depends on the directions of the fluids, that is, $\beta = 1$ for a counter-current heat exchanger whereas for a co-current one, $\beta = -1$. Equation (3) means that $T_h(l, t) = T_{h_l}(t)$ for a counter-current heat exchanger and $T_h(0, t) = T_{h_0}(t)$ for a co-current one. Thus, it allows to use a single model for both counter-current and co-current heat exchangers.

In the sequel, $L^2(\Omega)$ denotes the space of square-integrable functions (Atkinson and Han, 2009), defined on the domain Ω , and we assume that the state vector $T(z, t) = [T_h(z, t) T_c(z, t)]^T \in \mathcal{H}$ with

the state space $\mathcal{H} = L^2(\Omega) \oplus L^2(\Omega)$ endowed with the inner product

$$\langle f(z), g(z) \rangle_{\mathcal{H}} = \int_0^l f^T(z) g(z) dz, \quad \forall f(z) = \begin{bmatrix} f_1(z) \\ f_2(z) \end{bmatrix}, g(z) = \begin{bmatrix} g_1(z) \\ g_2(z) \end{bmatrix} \in \mathcal{H} \quad (7)$$

and the norm

$$\|f(z)\|_{\mathcal{H}}^2 = \langle f(z), f(z) \rangle_{\mathcal{H}} \quad (8)$$

The control problem of the heat exchanger consists in regulating the outlet temperature of the cold fluid towards a desired value. This objective can be achieved by manipulating either the inlet temperature or the flow rate of the hot fluid (Maidi et al., 2009). Nevertheless, controllability is a crucial question to be addressed before designing a controller for a heat exchanger because the reachable temperature depends on the inlet temperature and the flow rate, which are practically constrained (Alotaibi et al., 2004; Sano, 2007). As mentioned above, by manipulating the inlet temperature, the heat exchanger is a linear DPS and the controllability property can be assessed using some tools from the powerful semi-group theory (Sano, 2007). On other hand, by manipulating the flow rate, the heat exchanger is a bilinear DPS (a particular class of nonlinear systems) and the controllability analysis constitutes an open problem. The controllability of heat exchangers is less investigated in the literature. The conducted studies have shown that heat exchangers are controllable with the inlet temperatures and less controllable with flow rates (Alotaibi et al., 2004).

In the following, the studied problem is the control of the output $T_{c_l}(t)$ defined as the outlet temperature of the cold fluid, that is,

$$T_{c_l}(t) = T_c(l, t) \quad (9)$$

by manipulating the inlet temperature of the hot fluid, that is, by manipulating either $T_{h_0}(t)$ (case of the co-current) or $T_{h_l}(t)$ (case of the counter-current). Thus, the inlet temperature of the cold fluid $T_{c_0}(t)$ and both flow rates ϑ_h and ϑ_c are assumed to be disturbances.

Note that, according to the studied control problem, it follows that the actuator and the sensor are collocated (placed at the same boundary $z = 0$) in the case of co-current heat exchanger whereas for the counter-current heat exchanger they are anti-collocated (placed at opposite boundaries, i.e. $z = 0$

and $z = l$). In both cases, the heat exchanger is a linear DPS and the control problem can be tackled easily in the framework of geometric control. A summary of these characteristics is given in Table 1.

3 Geometric control of heat exchanger

Design of geometric control laws for DPSs is based on the concept of the characteristic index, which constitutes a generalization of the concept of relative order (Isidori, 1995; Kravaris and Kantor, 1990a) of a LPS. The characteristic index characterizes the spatio-temporal interactions between input and output variables of DPS (Christofides and Daoutidis, 1996). Mathematically, the characteristic index is the smallest order of the time derivative of the controlled variable which explicitly depends on the manipulated variable. A control geometric law can be designed if the characteristic index is finite. For DPSs, the characteristic index is finite if the manipulated variable (actuator) and the controlled variable (sensor) are collocated and is infinite if they are anti-collocated. This can be explained by the number of spatial points, assuming the idea of a spatial discretization, that the manipulated variable has to go through in order to observe its effect on the controlled output.

In this section, the geometric control laws that solve the formulated control problem in Section 2 for both counter-current and co-current heat exchangers are derived.

3.1 Control law design

Using the characteristic index, a control law that yields a linear DPS between an external input and the controlled output can be obtained. When the characteristic index is finite, two kinds of control law are possible: the *input-output linearizing* and the *output stabilizing* control laws, which yields an unstable (chain of integrators) and a stable linear LPS, respectively. These control laws are determined separately for the counter-current and co-current heat exchangers.

3.1.1 Counter-current heat exchanger

The characteristic index of a linear system is obtained by considering the successive time derivatives of the controlled output until the manipulated input appears explicitly. By setting $\beta = 1$ in the state

equation (2) and evaluating the first time derivative of the controlled output (9), we obtain

$$\frac{dT_{c_l}(t)}{dt} = \left. \frac{\partial T_c(z, t)}{\partial t} \right|_{z=l} \quad (10)$$

$$= \vartheta_c \left. \frac{\partial T_c(z, t)}{\partial z} \right|_{z=l} + \alpha_c (T_h(l, t) - T_c(l, t)) \quad (11)$$

$$= \vartheta_c \left. \frac{\partial T_c(z, t)}{\partial z} \right|_{z=l} + \alpha_c (T_{h_l}(t) - T_{c_l}(t)) \quad (12)$$

As the manipulated variable $T_{h_l}(t)$ appears explicitly in the first derivative of $T_{c_l}(t)$, the characteristic index is $\sigma = 1$. Note that, in the counter-current case, the manipulated and controlled variables are collocated. Solving the following equation

$$v(t) = \vartheta_c \left. \frac{\partial T_c(z, t)}{\partial z} \right|_{z=l} + \alpha_c (T_{h_l}(t) - T_{c_l}(t)) \quad (13)$$

with respect to the manipulated variable $T_{h_l}(t)$ yields the following *input-output linearizing* control law

$$T_{h_l}(t) = \frac{v(t) + \alpha_c T_{c_l}(t) - \vartheta_c \left. \frac{\partial T_c(z, t)}{\partial z} \right|_{z=l}}{\alpha_c} \quad (14)$$

where $v(t)$ is introduced as the external input.

Thus, in closed loop, the following linear LPS results

$$\frac{\bar{T}_{c_l}(s)}{V(s)} = \frac{1}{s} \quad (15)$$

where s is the Laplace variable, $\bar{T}_{c_l}(s)$ and $V(s)$ are the Laplace transforms of $T_{c_l}(t)$ and $v(t)$, respectively.

To obtain an *output stabilizing* control law, consider the following feedback control

$$v(t) = T_{c_l}(t) + \gamma_1 \frac{dT_{c_l}(t)}{dt} \quad (16)$$

where γ_1 is a time parameter chosen so that the resulting closed loop system is stable. The composition of Equations (9), (12) and (16) yields the following equation

$$v(t) = T_{c_l}(t) + \gamma_1 \vartheta_c \left. \frac{\partial T_c(z, t)}{\partial z} \right|_{z=l} + \gamma_1 \alpha_c (T_{h_l}(t) - T_{c_l}(t)) \quad (17)$$

and by solving the resulting equation with respect to the manipulated variable $T_{h_l}(t)$, the following *output stabilizing* feedback results

$$T_{h_l}(t) = \frac{v(t) + (\gamma_1 \alpha_c - 1) T_{c_l}(t) - \gamma_1 \vartheta_c \left. \frac{\partial T_c(z, t)}{\partial z} \right|_{z=l}}{\gamma_1 \alpha_c} \quad (18)$$

which yields the following closed loop first order transfer function

$$\frac{\bar{T}_{c_l}(s)}{V(s)} = \frac{1}{\gamma_1 s + 1} \quad (19)$$

3.1.2 Co-current heat exchanger

In the case of the co-current heat exchanger, the manipulated and controlled variables are non-collocated, opposite to the counter-current heat exchanger where they were collocated, thus the characteristic index is infinite (Maidi et al., 2010). To design the geometric control law, we use the approach proposed by Maidi and Corriou (2011a) that consists in the following three steps:

1. Define an auxiliary controlled output $T_c^m(t)$ as the spatial weighed average temperature of the cold fluid, that is,

$$T_c^m(t) = \mathcal{C} T(z, t) \quad (20)$$

$$= \int_0^l [0 \quad c(z)] T(z, t) dz \quad (21)$$

$$= \int_0^l c(z) T_c(z, t) dz \quad (22)$$

where $\mathcal{C}$ is a linear bounded operator and $c(z)$ is a smooth shaping function that satisfies $c(0) \neq 0$.

2. Insert the manipulated variable $T_{h_0}(t)$ in the state equation (1) using the notion of the extended operator (Stafford and Dowrick, 1977). It allows to convert the boundary control problem into a punctual one.
3. Derive the geometric feedback to control the auxiliary controlled output $T_c^m(t)$ using the equivalent punctual control model obtained at Step 2.
4. Define the desired reference $v(t)$ of the auxiliary controlled output $T_c^m(t)$ using either a controller

or the steady-state mapping between the controlled variable $T_{c_l}(t)$ and the auxiliary output $T_c^m(t)$.

Hence, in the case of the co-current heat exchanger, the following equivalent distributed control model can be obtained using the Laplace transform as proposed by Maidi and Corriou (2011a):

$$\frac{\partial T_h(z, t)}{\partial t} = -\vartheta_h \frac{\partial T_h(z, t)}{\partial z} + \alpha_h (T_c(z, t) - T_h(z, t)) + \vartheta_h \delta_\varepsilon(z) T_{h_0}(t) \quad (23)$$

$$\frac{\partial T_c(z, t)}{\partial t} = -\vartheta_c \frac{\partial T_c(z, t)}{\partial z} + \alpha_c (T_h(z, t) - T_c(z, t)) \quad (24)$$

with the following new boundary conditions

$$T_h(0, t) = 0 \quad (25)$$

$$T_c(0, t) = T_{c_0}(t) \quad (26)$$

where $T_{h_0}(t)$ is assumed to be spanned over a small portion of the spatial domain Ω (close to the boundary $z = 0$), that is,

$$\delta_\varepsilon(z) = \begin{cases} \frac{1}{\varepsilon} & \text{for } 0 \leq z \leq \varepsilon \\ 0 & \text{Otherwise} \end{cases} \quad (27)$$

Thus, the evaluation of the first time derivative of the auxiliary output (22) gives

$$\frac{dT_c^m(t)}{dt} = \int_0^l c(z) \frac{\partial T_c(z, t)}{\partial t} dz \quad (28)$$

$$= \int_0^l c(z) \left[-\vartheta_c \frac{\partial T_c(z, t)}{\partial z} + \alpha_c (T_h(z, t) - T_c(z, t)) \right] dz \quad (29)$$

The first time derivative (29) does not depend explicitly on the manipulated variable $T_{h_0}(t)$, conse-

quently the characteristic index σ is greater than one. Performing a second time derivative, we obtain

$$\frac{d^2 T_c^m(t)}{dt^2} = \int_0^l c(z) \frac{\partial}{\partial t} \left[-\vartheta_c \frac{\partial T_c(z, t)}{\partial z} + \alpha_c (T_h(z, t) - T_c(z, t)) \right] dz \quad (30)$$

$$= \int_0^l c(z) \left[\vartheta_c^2 \frac{\partial^2 T_c(z, t)}{\partial z^2} + 2 \vartheta_c \alpha_c \frac{\partial T_c(z, t)}{\partial z} - \alpha_c (\vartheta_c + \vartheta_h) \frac{\partial T_h(z, t)}{\partial z} \right. \\ \left. + \alpha_c (\alpha_h + \alpha_c) (T_c(z, t) - T_h(z, t)) + \alpha_c \vartheta_h \delta_\varepsilon(z) T_{h_0}(t) \right] dz \quad (31)$$

We have

$$\int_0^l c(z) \delta_\varepsilon(z) dz = c(0) \neq 0 \quad (32)$$

hence

$$\frac{d^2 T_c^m(t)}{dt^2} = \int_0^l c(z) \left[\vartheta_c^2 \frac{\partial^2 T_c(z, t)}{\partial z^2} + 2 \vartheta_c \alpha_c \frac{\partial T_c(z, t)}{\partial z} - \alpha_c (\vartheta_c + \vartheta_h) \frac{\partial T_h(z, t)}{\partial z} \right. \\ \left. + \alpha_c (\alpha_h + \alpha_c) (T_c(z, t) - T_h(z, t)) \right] dz + \alpha_c \vartheta_h c(0) T_{h_0}(t) \quad (33)$$

Notice that the manipulated variable appears explicitly in the second derivative of $T_c^m(t)$, consequently the characteristic index is $\sigma = 2$. By solving the following equation expressing the external input

$$v(t) = \int_0^l c(z) \left[\vartheta_c^2 \frac{\partial^2 T_c(z, t)}{\partial z^2} + 2 \vartheta_c \alpha_c \frac{\partial T_c(z, t)}{\partial z} - \alpha_c (\vartheta_c + \vartheta_h) \frac{\partial T_h(z, t)}{\partial z} \right. \\ \left. + \alpha_c (\alpha_h + \alpha_c) (T_c(z, t) - T_h(z, t)) \right] dz + \alpha_c \vartheta_h c(0) T_{h_0}(t) \quad (34)$$

with respect to the manipulated variable $T_{h_0}(t)$, we obtain the following *input-output linearizing* feedback

$$T_{h_0}(t) = \frac{v(t) - \int_0^l c(z) \left[a \frac{\partial^2 T_c(z, t)}{\partial z^2} + b \frac{\partial T_c(z, t)}{\partial z} - c \frac{\partial T_h(z, t)}{\partial z} + d \Delta T(z, t) \right] dz}{\alpha_c \vartheta_h c(0)} \quad (35)$$

with $a = \vartheta_c^2$, $b = 2 \vartheta_c \alpha_c$, $c = \alpha_c (\vartheta_c + \vartheta_h)$, $d = \alpha_c (\alpha_h + \alpha_c)$ and $\Delta T(z, t) = T_c(z, t) - T_h(z, t)$.

The state feedback (35) yields the following closed loop system:

$$\frac{\bar{T}_c^m(s)}{V(s)} = \frac{1}{s^2} \quad (36)$$

To obtain the *output stabilizing* feedback, as the characteristic index $\sigma = 2$, we consider the following feedback

$$v(t) = T_c^m(t) + \gamma_1 \frac{dT_c^m(t)}{dt} + \gamma_2 \frac{d^2T_c^m(t)}{dt^2} \quad (37)$$

Combining Equations (29), (33) and (37), the following output stabilizing feedback follows

$$T_{h_0}(t) = \frac{v(t) - T_c^m(t) - \int_0^t c(z) \left[a \frac{\partial^2 T_c(z, t)}{\partial z^2} + b \frac{\partial T_c(z, t)}{\partial z} - c \frac{\partial T_h(z, t)}{\partial z} + d \Delta T(z, t) \right] dz}{\gamma_2 \alpha_c \vartheta_h c(0)} \quad (38)$$

with $a = \gamma_2 \vartheta_c^2$, $b = \vartheta_c (2 \gamma_2 \alpha_c - \gamma_1)$, $c = \gamma_2 \alpha_c (\vartheta_c + \vartheta_h)$ and $d = \alpha_c (\gamma_2 (\alpha_h + \alpha_c) - \gamma_1)$. The state feedback (38) yields the following closed loop second order transfer function

$$\frac{\bar{T}_c^m(s)}{V(s)} = \frac{1}{\gamma_2 s^2 + \gamma_1 s + 1} \quad (39)$$

In the following, we focus on the output stabilizing state feedbacks, which are more practical because they yield a closed loop system, which is externally stable.

3.2 Stability of the closed loop

The stability of transfer functions (19) and (39), by an adequate choice of the tuning parameters γ_1 and γ_2 , is not sufficient to ensure the stability of the resulting closed loop system because there is an internal dynamics related to the unobservable part of the controlled outputs. In the following, we demonstrate that, as the heat exchanger is a minimum-phase system, the internal dynamics of the resulting closed loop system is exponentially stable. In the following, we denote by $u(t)$ and $y(t)$ the manipulated and controlled variables, respectively. Thus, in the case of counter-current heat exchanger, $u(t) = T_{h_i}(t)$ and $y(t) = T_{c_i}(t)$ while in the case of co-current heat exchanger $u(t) = T_{h_0}(t)$ and $y(t) = T_c^m(t)$.

Noting

$$T(z, t) = \begin{bmatrix} T_h(z, t) \\ T_c(z, t) \end{bmatrix} \quad (40)$$

the models of counter-current and co-current heat exchangers can be written under the following abstract form (Curtain and Zwart, 1995)

$$\frac{\partial T(z, t)}{\partial t} = \mathcal{A}T(z, t) + \mathcal{B}u(t) \quad (41)$$

where the state operator $\mathcal{A}$ is defined as follows

$$\mathcal{A} = \begin{bmatrix} -\vartheta_h & 0 \\ 0 & \beta \vartheta_c \end{bmatrix} \frac{\partial}{\partial z} + \begin{bmatrix} -\alpha_h & \alpha_h \\ \alpha_c & -\alpha_c \end{bmatrix} \quad (42)$$

and the control operator $\mathcal{B}$ is given as

$$\mathcal{B} = \begin{bmatrix} \vartheta_h \left(\frac{(1-\beta)}{2} \delta_\varepsilon(z) - \frac{(1+\beta)}{2} \delta_\varepsilon(l-z) \right) \\ 0 \end{bmatrix} \quad (43)$$

with

$$\delta_\varepsilon(l-z) = \begin{cases} \frac{1}{\varepsilon} & \text{for } l-\varepsilon \leq z \leq l \\ 0 & \text{Otherwise} \end{cases} \quad (44)$$

and $\delta_\varepsilon(z)$ is given by (27).

Both state feedbacks (18) and (38) can be written under the following form

$$u(t) = \mathcal{K}T(z, t) + \mathcal{L}e(t) \quad (45)$$

where $\mathcal{K}$ and $\mathcal{L}$ are spatial operators that can be easily identified from the state feedback, and $e(t)$ denotes the tracking error, that is, $e(t) = v(t) - y(t)$.

Substituting the state feedback (45) into (41), the following closed loop system results

$$\frac{\partial T(z, t)}{\partial t} = \mathcal{A} T(z, t) + \mathcal{B} (\mathcal{K} T(z, t) + \mathcal{L} e(t)) \quad (46)$$

$$T_h \left(\frac{(1 + \beta)}{2} l, t \right) = 0 \quad (47)$$

$$T_c(0, t) = T_{c_0}(t) \quad (48)$$

$$T_h(z, 0) = T_h^*(z) \quad (49)$$

$$T_c(z, 0) = T_c^*(z) \quad (50)$$

and can be written as the interconnection of two subsystems as follows

$$\frac{dy(t)}{dt} = -\frac{1}{\gamma_1} y(t) + \frac{1}{\gamma_1} v(t) \quad (51)$$

$$\frac{\partial T(z, t)}{\partial t} = (\mathcal{A} + \mathcal{B} \mathcal{K}) T(z, t) + \mathcal{B} \mathcal{L} e(t) \quad (52)$$

$$T_h \left(\frac{(1 + \beta)}{2} l, t \right) = 0 \quad (53)$$

$$T_c(0, t) = T_{c_0}(t) \quad (54)$$

$$T_h(z, 0) = T_h^*(z) \quad (55)$$

$$T_c(z, 0) = T_c^*(z) \quad (56)$$

By zeroing the auxiliary output ($T_c^m(t) = 0$) (Christofides and Daoutidis, 1996), the following zero dynamics results

$$\frac{\partial T(z, t)}{\partial t} = (\mathcal{A} + \mathcal{B} \mathcal{K}) T(z, t) \quad (57)$$

$$T_h \left(\frac{(1 + \beta)}{2} l, t \right) = 0 \quad (58)$$

$$T_c(0, t) = T_{c_0}(t) \quad (59)$$

$$T_h(z, 0) = T_h^*(z) \quad (60)$$

$$T_c(z, 0) = T_c^*(z) \quad (61)$$

Now, as the counter-current heat exchanger is a minimum-phase system, we conclude that the operator $\mathcal{A} + \mathcal{BK}$ generates a stable C_0 -semigroup $\mathcal{U}(t)$ (Christofides and Daoutidis, 1996), consequently (Pazy, 1983)

$$\|\mathcal{U}(t)\|_{\mathcal{H}} \leq M_T e^{-\omega_T t}, \quad M_T \geq 1, \omega_T > 0 \quad (62)$$

As the state feedback (45) yields a stable $v(t)$ - $y(t)$ system, hence for a bounded external input $v(t)$, we have

$$|e(t)| \leq M_e e^{-\omega_e t}, \quad M_e \geq 1, \omega_e > 0 \quad (63)$$

where $|\cdot|$ is the Euclidean norm.

Thus, from the C_0 -semigroup theory, as the tracking error $e(t) \in L^1(\mathbb{R}^+)$, the internal dynamics admits the following mild solution (Curtain and Zwart, 1995)

$$T(z, t) = \mathcal{U}(t) T(z, 0) + \int_0^t \mathcal{U}(t - \zeta) \mathcal{BL} e(\zeta) d\zeta \quad (64)$$

Thereafter

$$\|T(z, t)\|_{\mathcal{H}} = \left\| \mathcal{U}(t) T(z, 0) + \int_0^t \mathcal{U}(t - \zeta) \mathcal{BL} e(\zeta) d\zeta \right\|_{\mathcal{H}} \quad (65)$$

$$\leq \|T(z, 0)\|_{\mathcal{H}} \|\mathcal{U}(t)\|_{\mathcal{H}} + \left\| \int_0^t \mathcal{U}(t - \zeta) \mathcal{BL} e(\zeta) d\zeta \right\|_{\mathcal{H}} \quad (66)$$

$$\leq \|T(z, 0)\|_{\mathcal{H}} M_T e^{-\omega_T t} + \int_0^t M_T e^{-\omega_T (t-\zeta)} \|\mathcal{BL}\|_{\mathcal{H}} M_e e^{-\omega_e \zeta} d\zeta \quad (67)$$

$$\leq \|T(z, 0)\|_{\mathcal{H}} M_T e^{-\omega_T t} + M_T M_e \|\mathcal{BL}\|_{\mathcal{H}} e^{-\omega_T t} \int_0^t e^{(\omega_T - \omega_e) \zeta} d\zeta \quad (68)$$

Now, in the case of $\omega_x = \omega_e$, it follows that

$$\|T(z, t)\|_{\mathcal{H}} \leq M_T \left(\|T(z, 0)\|_{\mathcal{H}} + M_e \|\mathcal{BL}\|_{\mathcal{H}} t \right) e^{-\omega_T t} \quad (69)$$

and if $\omega_x \neq \omega_e$, we have

$$\|T(z, t)\|_{\mathcal{H}} \leq \|T(z, 0)\|_{\mathcal{H}} M_T e^{-\omega_T t} + \frac{M_T M_e \|\mathcal{BL}\|_{\mathcal{H}}}{\omega_T - \omega_e} (e^{-\omega_e t} - e^{-\omega_T t}) \quad (70)$$

$$\leq M_T \left(\|T(z, 0)\|_{\mathcal{H}} + \frac{M_e \|\mathcal{BL}\|_{\mathcal{H}}}{|\omega_T - \omega_e|} \right) e^{-\omega t} \quad (71)$$

where $\omega = \min\{\omega_T, \omega_e\}$.

Hence, from relations (69) and (71), it can be concluded that the internal dynamics of the closed loop system is exponentially stable.

3.3 Practical implementation issues

3.3.1 State estimation

The control laws presented in Subsection 3.1 are of distributed nature (infinite dimensional), that is, for practical implementation, the whole state must be available. Practically, due to its infinite dimension, the whole state cannot be measured. Consequently, an observer must be designed to reconstruct the entire state from the accessible measurements (Hidayat et al., 2011). Before attempting to design an observer for a heat exchanger, the analysis of the observability property is the primary question. This property depends on the number of sensors and their locations (Ray, 1989; Waldraff et al., 1998). Observability can be analyzed using either the approximate model (indirect method) or the PDEs model (direct method) (Ray, 1989; Sano, 2007; Singh and Hahn, 2007). Note that indirect methods can mask this property and must be applied with care because they may lead to erroneous conclusions (Ray, 1989; Singh and Hahn, 2007). For the counter-current and co-current heat exchangers, both inlet and outlet temperatures are accessible for measurement. According to the studies conducted in the literature, using the PDEs model (direct approach), it is shown that, by sensing these temperatures, the observability condition is satisfied (Sano, 2007).

For observer design, both direct and indirect approaches can be used (Hidayat et al., 2011). Note that even if the direct approach using the PDE model is recommended, an approximation method must be used for practical implementation. Nevertheless, attention must be paid to the approximation method used so that the observability condition must be preserved (Ray, 1989; Singh and Hahn, 2007). In this chapter, the Kalman observer is used to reconstruct the entire temperatures of both cold and hot fluids. Thus, using the method of lines (Van de Wouwer et al., 2004), based on the finite difference scheme, and by considering N discretization points, the infinite dimensional state equations (1) and (2) of the

heat exchanger can be approximated by the following finite dimensional state equation

$$\dot{T}_\Delta(t) = A T_\Delta(t) + B u(t) + E T_{c_0}(t) \quad (72)$$

$$y(t) = C x(t) \quad (73)$$

where the state vector is defined according to the type of the heat exchanger as follows:

- For the counter-current heat exchanger:

$$T_\Delta(t) = [T_{h_0}(t) T_{h_2}(t) \dots T_{h_{N-1}}(t) T_{c_1}(t) T_{c_2}(t) \dots T_{c_N}(t)]^T \quad (74)$$

- For the co-current heat exchanger:

$$T_\Delta(t) = [T_{h_1}(t) T_{h_2}(t) \dots T_{h_N}(t) T_{c_1}(t) T_{c_2}(t) \dots T_{c_N}(t)]^T \quad (75)$$

where $T_{h_i}(t)$ and $T_{c_i}(t)$ ($i = 1, \dots, N$) are the temperatures of the hot and cold fluids temperatures at the position $z = i \Delta z$ with $\Delta z = l/N$ the discretization step. $A \in \mathbb{R}^{2N \times 2N}$, $B \in \mathbb{R}^{2N \times 1}$, $E \in \mathbb{R}^{2N \times 1}$ and $C \in \mathbb{R}^{1 \times 2N}$ are the state, control, disturbance and observation matrices, respectively. These matrices depend both on the used finite difference scheme and the type of the heat exchanger. More details about the discretization of both heat exchangers can be found in Corriou (2018).

In the presence of both process and measurement noises, the approximate model of the heat exchanger is given as follows

$$\dot{T}_\Delta(t) = A T_\Delta(t) + B u(t) + E T_{c_0}(t) + \zeta(t) \quad (76)$$

$$y(t) = C T_\Delta(t) + \eta(t) \quad (77)$$

where $\zeta(t)$ is the process noise and $\eta(t)$ is the measurement noises with zero mean. Q and R are the covariance of $\zeta(t)$ and $\eta(t)$, respectively.

As the heat exchanger is continuous and the measurements are in discrete form, we use the discrete-

continuous (Corriou, 2018) formulation of the Kalman filter

$$\dot{\hat{T}}_{\Delta}(t) = A \hat{T}_{\Delta}(t) + B u(t) + E T_{c_0}(t) \quad (78)$$

$$\dot{P}(t) = A P(t) + P(t) A^T + Q \quad (79)$$

where $P(t)$ is the covariance matrix.

The estimate $\hat{T}_{\Delta}(t)$ of the state $T_{\Delta}(t)$ from the available measurements $T_{c_0}(t)$ (measured disturbance) and $y(t)$ is provided by the Kalman filter in two phases (Corriou, 2018):

Prediction phase: The equations (78) and (79) are integrated from time $t = k - 1$ to $t = k$ to obtain the estimate $\hat{T}_{\Delta}(k|k - 1)$ and $\hat{P}(k|k - 1)$. $\hat{T}_{\Delta}(k|k - 1)$ being the estimate of $T_{\Delta}(t)$ at $t = k$ from the measurement at $t = k - 1$.

Correction phase: The estimate obtained in the prediction phase for $t = k$ is corrected and the covariance matrix $P(t)$ is updated as follows:

$$\hat{T}_{\Delta}(k|k) = \hat{T}_{\Delta}(k|k - 1) + P(k|k - 1) C^T (C P(k|k - 1) C^T + R)^{-1} (y(k) - C \hat{T}_{\Delta}(k|k - 1)) \quad (80)$$

$$P(k|k) = (I - K_k C) P(k|k - 1) \quad (81)$$

Note that, in the case of co-current heat exchanger, the auxiliary controlled output $T_c^m(t)$, given by Equation (22), can be evaluated at $t = k$ as follows

$$T_c^m(t) = \mathcal{C}_{\Delta} \hat{T}_{\Delta}(t) \quad (82)$$

with $\hat{T}_{\Delta}(t) \equiv \hat{T}_{\Delta}(k|k)$ and the row vector $\mathcal{C}_{\Delta}$ depends on the quadrature used to approximate the integral operator $\mathcal{C}$ according to (Equation (22)).

3.3.2 Robustness

The control laws designed in Subsection 3.1 for both heat exchangers are sensitive to modeling errors and parameter uncertainties. For robustness and for stabilization, in the case of input-output linearizing control laws (14) and (35), we use the global linearizing control structure (Corriou, 2018; Kravaris and

Kantor, 1990b) where the external input $v(t)$ is defined by means of a robust controller as follows:

- Case of the counter-current heat exchanger:

$$v(t) = \int_0^t G_c(t - \xi) (T_{c_l}^d(\xi) - T_{c_l}(\xi)) d\xi \quad (83)$$

- Case of the co-current heat exchanger:

$$v(t) = \int_0^t G_c(t - \xi) (T_c^{m^d}(\xi) - T_c^m(\xi)) ds \quad (84)$$

where $T_{c_l}^d(t)$ and $T_c^{m^d}(t)$ are desired references for $T_{c_l}(t)$ and $T_c^m(t)$, respectively, and the kernel $G_c(t)$ characterizes the desirable dynamical behavior for the overall control system.

Recall that, in the case of the co-current heat exchanger, the developed control laws (35) and (38) force the auxiliary output to track the desired reference $T_c^{m^d}(t)$. To solve the initial control problem formulated in Section 2, we propose to define the desired reference $T_c^{m^d}(t)$ by an external robust controller as follows

$$T_c^{m^d}(t) = \int_0^t G_c(t - \xi) (T_{c_l}^d(\xi) - T_{c_l}(\xi)) d\xi \quad (85)$$

The control strategies for the counter-current and co-current heat exchangers are summarized in Figures 2 and 3.

4 Closed-loop performance evaluation

In this Section, the performances of the global linearizing control structures based on the output stabilizing state feedback are evaluated through numerical simulation for both counter-current and co-current heat exchangers. Both the external input $v(t)$ and the desired reference $T_c^{m^d}(t)$ are defined by means of a PI controller, that is,

$$G_c(s) = k_c \left(1 + \frac{1}{\tau_i s} \right) \quad (86)$$

and their parameters k_c and τ_i are tuned using the Graham and Lathrop method that minimizes the integral of time-weighted absolute error (ITAE) criterion (Corriou, 2018).

The closed loop system is simulated using the method of lines (Van de Wouwer et al., 2004), based

on the finite difference scheme, by considering $N = 100$ discretization points. For state estimation, the Kalman filter is designed using the approximate LPS model obtained by the method of lines. The integral terms involved in the control laws are evaluated using the trapezoidal quadrature. The different parameters used in simulation runs are summarized in Table 2.

The initial conditions $T_h^*(z)$ and $T_c^*(z)$ are the steady-state spatial temperature profiles obtained for an inlet cold fluid temperature equal to 25°C and inlet hot fluid temperature equal to 50°C. The used measurements $T_{c_0}(t)$ and $y(t)$ are assumed to be corrupted with Gaussian white noise signals of standard deviation equal to 0.2°C.

Both output tracking and disturbance rejection performances have been evaluated. The inlet temperature of the cold fluid $T_{c_0}(t)$ is assumed as a disturbance. The simulation test consists in applying two step changes $T_c^d(t) = 120^\circ\text{C}$ and $T_c^d(t) = 60^\circ\text{C}$ at $t = 1\text{ s}$ and $t = 10\text{ s}$, respectively for the counter-current heat exchanger while for the co-current heat exchanger they are applied at $t = 1\text{ s}$ and $t = 40\text{ s}$, respectively. Between these steps changes, a step of 100% and of -10% of the temperature of the inlet cold fluid $T_{c_0}(t)$ is imposed as an external disturbance at $t = 5\text{ s}$ and at $t = 20\text{ s}$ for the counter-current and co-current heat exchangers, respectively. To avoid sudden fluctuations of the manipulated variable due to the brutal set points change, the imposed set point is smoothed by a first order filter with a time constant equal to 0.2 s.

In the simulation runs, the effect of the measurements errors on the quality of the estimation is minimized by a simple filtering using a moving average filter with a forgetting factor equal to 0.96 (Corriou, 2018). Also, the manipulated variable is held constant over a sampling period $\Delta t = 0.02\text{ s}$.

The obtained results are depicted in Figures 4–5 and Figures 6–7 for the counter-current and co-current heat exchangers, respectively. It can be seen that the output tracking is achieved in a satisfactory manner. But in the case of co-current heat exchanger, the controlled output $T_{c_i}(t)$ presents slight fluctuations around the imposed set points. This is due of course to the effect of the noise measurements. Indeed, by comparing the quality of the control moves given by Figures 5 and 7, it follows that in the case of the co-current heat exchanger, the quality of the control is more sensitive to noisy measurements. This is expected because the state feedback (38) is more sensitive to the measurement noise since it involves more spatial derivatives compared to the state feedback (18). Note that without noise measurements, perfect tracking is achieved for both heat exchangers.

Also, from the results obtained, it can be observed that the disturbance effect is damped, which

demonstrates the disturbance rejection capability of both control strategies. Notice that, in the case of the counter-current heat exchanger (collocated case), the effect of the disturbance can be observed on the controlled output only if the disturbance is important. On another side, the disturbance effect is attenuated as the observation position moves and the effect becomes nearly invisible at the outlet of the heat exchanger (Maidi et al., 2009).

5 Conclusion

Based on the geometric control theory, a state feedback of infinite dimensional nature is designed both for counter-current and co-current heat exchangers to control the outlet temperature of the cold fluid by manipulating the inlet temperature of the hot fluid. The design approach is based on the PDE model. Two kinds of state feedback are developed: input-output linearizing and output stabilizing control laws. In the case of the counter-current heat exchanger, it is shown that, as the actuator and the sensor are collocated, the characteristic index is finite and the control laws can be designed easily by evaluating the first time derivative of the outlet. In the case of the co-current heat exchanger, the characteristic index is infinite because the sensor and the actuator are non-collocated, so that a different design approach consisting in four steps is presented. The idea consists in introducing the manipulated variable in the state equation using the notion of the extended operator, and then a control law is designed by assuming an auxiliary output, defined as the spatial average temperature of the cold temperature, instead of its outlet temperature. To achieve the original control problem, a control strategy is presented where the desired reference for the auxiliary output is defined by means of an external controller. The robustness, the stability of the resulting closed loop systems and the implementation of the developed control laws issues are discussed. The tracking and disturbance rejection performances are evaluated via numerical simulation for both counter-current and co-current heat exchangers.

Notice that the direct design approaches, presented in this chapter, can be adopted to design infinite dimensional boundary controllers for 2×2 hyperbolic systems of balance laws.

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	Manipulated variable	Controlled variable	Actuator and sensor location
Counter-current	$T_{h_i}(t)$	$T_{c_l}(t)$	Collocation
Co-current	$T_{h_0}(t)$	$T_{c_l}(t)$	Anti-collocation

Table 1: Manipulated and controlled variables for the heat exchangers.

Figure 1: Tube heat exchangers.

Figure 2: Global control strategy for counter-current heat exchanger.

Figure 3: Global control strategy for co-current heat exchanger.

	Counter-current heat exchanger	Co-current heat exchanger
Physical parameters	$v_c = 1 \text{ m} \cdot \text{s}^{-1}, v_h = 2 \text{ m} \cdot \text{s}^{-1}, \alpha_c = 1 \text{ s}^{-1}, \alpha_h = 1 \text{ s}^{-1}$	
Initial inlet temperature ($t = 0$)	$T_{c_0} = 25^\circ\text{C}, T_{h_i} = 50^\circ\text{C}$	$T_{c_0} = 25^\circ\text{C}, T_{h_0} = 50^\circ\text{C}$
State feedback	$\gamma_1 = 1 \text{ s}$	$\gamma_1 = 0.22 \text{ s}, \gamma_2 = 5 \times 10^{-2} \text{ s}^2$
Robust controller	$k_c = 7.8, \tau_i = 1.03 \text{ s}$	$k_c = 2.4 \times 10^{-2}, \tau_i = 4.69 \times 10^{-2} \text{ s}$
External controller	—	$k_c = 2.68 \times 10^{-2}, \tau_i = 0.3253 \text{ s}$
Kalman filter	$P(0/0) = 10^{-2} I_{2N \times 2N}, Q = 5 \times 10^{-2} I_{2N \times 2N}, R = 10^{-2}$	$Q = P(0/0) = 0.1 I_{2N \times 2N}, R = 0.1 I_{2 \times 2}$

Table 2: Parameters used in simulation.

Figure 4: Counter-current heat exchanger: Evolution of the controlled output $T_{c_l}(t)$.

Figure 5: Counter-current heat exchanger: Evolution of the manipulated variable $T_{h_l}(t)$

Figure 6: Co-current heat exchanger: Evolution of the controlled output $T_{c_l}(t)$.

Figure 7: Co-current heat exchanger: Evolution of the manipulated variable $T_{h_0}(t)$.