

Dissipative boundary PI controller for an adiabatic plug-flow reactor with mass recycle

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Mathematical Systems Theory

Outline

- 1 Motivation
 - Introduction
 - Convective Distributed Parameter System with mass recycle
- 2 Dissipative properties and boundary control
 - Dissipative properties
 - Boundary control
- 3 Case study
 - Mass and energy balances
 - Numerical simulations



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Introduction

- Dissipation, storage function and supply rate (Willems, 1972).
- Stability analysis based on the internal entropy production (García-Sandoval et al., 2016).
- Tubular reactors with mass recycle (Reilly and Schmitz, 1966).
- Using thermodynamic properties to regulate nonlinear systems such as plug flow reactors is still an open field.



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System description

- Convective Distributed Parameter System

$$\frac{\partial \eta(z, t)}{\partial t} = -v \frac{\partial \eta(z, t)}{\partial z} + Mf(\zeta), \quad (1)$$

where: state vector $\eta(z, t) \in \mathcal{H}^n[(0, L), \mathbb{R}^n]$ (extensive properties) and are the respective entropy-conjugated co-states $\zeta(z, t) \in \mathcal{H}^n[(0, L), \mathbb{R}^n]$ (intensive properties).
 $z \in [0, L] \subset \mathbb{R}$, $t \in [0, \infty]$, $v = \frac{F}{A} \in \mathbb{R}^+$.

- Initial conditions:

$$\eta(z, 0) = \alpha(z), \quad (2)$$

where $\alpha(z) \in \mathcal{H}^n[(0, L), \mathbb{R}^n]$.

- Boundary condition with mass recycle:

$$\eta(0, t) = (1 - u(t)) \eta_{in}(t) + u(t) \eta(L, t). \quad (3)$$



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Entropy balance

- Entropy as a state variable: $s = \phi(\boldsymbol{\eta})$.
- Entropy gradient:

$$\zeta(z, t) := \nabla s(z, t), \quad (4)$$

- Given the Gibbs relation: $ds = \zeta^T d\boldsymbol{\eta}$, the entropy dynamics is:

$$\frac{\partial s(z, t)}{\partial t} = -v \frac{\partial s(z, t)}{\partial z} + \sigma(z, t), \quad (5)$$

where $\sigma(z, t) := \zeta^T Mf(\zeta)$ is the internal entropy production density.

- Total entropy: $S(t) = A \int_0^L s(\lambda, t) d\lambda$, with dynamics

$$\frac{dS(t)}{dt} = -F [s(L, t) - s(0, t)] + \Sigma(t), \quad (6)$$

where $\Sigma(t) = A \int_0^L \sigma(\lambda, t) d\lambda \geq 0$.



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Entropy as storage function

Proposition

Consider a convective thermodynamic system as the previously described with constant velocities (Eqs. (1)-(3)). If the system is thermodynamically consistent, then it is dissipative with the entropy as the storage function and the internal entropy production as the dissipation.

Proof.

(Sketch) Let us consider $W(t) := \bar{S} - S(t)$, then

$$\frac{dW(t)}{dt} = F [s(L, t) - s(0, t)] - \Sigma(t). \text{ Since } \Sigma(t) \geq 0,$$

$$W(t) - W(0) \leq \int_0^t \omega(\eta_0(\theta), \eta_L(\theta)) d\theta, \text{ with}$$

$$\omega(\eta_0(t), \eta_L(t)) := F [s_L(t) - s_0(t)] = F [s(L, t) - s(0, t)]. \quad \square$$



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Boundary control definition

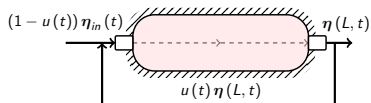


Figure: System with mass recycle.

- Boundary condition:

$$\eta(0, t) = (1 - u(t)) \mathbf{d}(t) + u(t) \eta(L, t).$$

- Disturbance vector:
 $\mathbf{d}(t) := \eta_{in}(t),$
- Regulated variable:
 $y(t) := \eta_i(L, t),$
- Control variable: $u(t)$
 (recycle rate).



Controller structure

General controller

A controller that solve the problem is:

$$\dot{u}(t) = (\dot{\chi}_{ref}(t) - \kappa_1 e_1 - w(t)) / \Phi(t),$$

where $e_1(t) = \chi(t) - \chi_{ref}(t)$, $\Phi(t) := F\zeta_0^T(t) [\eta_L(t) - \mathbf{d}(t)]$,

$\zeta_0(t) = \zeta(0, t)$ and $\eta_L(t) = \eta(0, t)$, while $\dot{\chi}_{ref}(t)$ is the reference of $\chi(t) := dW/dt$ (storage function dynamics). Finally,

$$w(t) := \frac{d^2W}{dt^2} + F\zeta_0^T(t) [\eta_L(t) - \mathbf{d}(t)] \dot{u}(t).$$

Particular controller

Considering a PI structure for $\dot{\chi}_{ref}(t)$ the controller reduce to

$$\dot{u}(t) = \left(K_p e_2(t) + \frac{K_p}{\tau_I} \int_0^t e_2(\tau) d\tau \right) / \Phi(t),$$

where, $e_2 = y(t) - y_{ref}$, while K_p and τ_I are the tuning parameters.



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Model: Single reversible reaction: $E_1 \rightleftharpoons 2E_2$

- Mass and internal energy balances:

$$\frac{\partial C_i(z, t)}{\partial t} = -v \frac{\partial C_i}{\partial z} + \Gamma_i r, \quad i = 1, 2$$

$$\frac{\partial U(z, t)}{\partial t} = -v \frac{\partial U}{\partial z}$$

where $\mathbf{C}^T(z, t) = (C_1 \ C_2)$, $U(z, t) = \mathbf{C}^T \mathbf{u}$,

$\mathbf{u} = \mathbf{u}_{ref} + \int_{T_{ref}}^T \mathbf{c}_v dT$, $r(z, t) = k (X_1 - K_{eq} X_2^2)$ with

$k = D_R \exp\left(-\frac{\varepsilon_0(\Gamma^T \mu^0) + E_{a0}}{RT}\right)$ and $K_{eq} = \exp\left(\frac{\Gamma^T \mu^0}{RT}\right)$.

- Boundary conditions of the system are

$$\mathbf{C}(0, t) = (1 - u(t)) \mathbf{d}_1(t) + u(t) \mathbf{C}(L, t), \quad (7)$$

$$U(0, t) = (1 - u(t)) d_2(t) + u(t) U(L, t). \quad (8)$$



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Parameters and space-time variation

Parameter	Value
R (J/mol·K)	8.314
s_{ref}^T/R (-)	(30.06 10.82)
u_{ref}^T/R (K)	$(1.623 \ 0.481) \times 10^4$
c_p^T/R (-)	(21.65 10.22)
ε_0 (-)	0.5
E_{a0} (J/mol)	2.2×10^5
D_R (1/h)	1.0×10^{20}
K_p/F (h/m ³)	1.5×10^4
$1/\tau_I F$ (1/m ³)	0.18

Table: Thermodynamic properties and parameters for the numerical

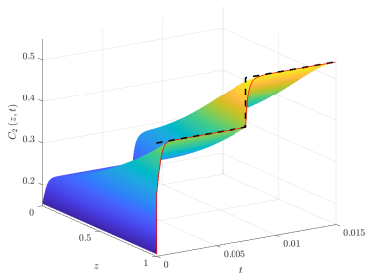


Figure: $C_{2,L}(t)$ regulation with a set point change at $t = t_{end}/2$.



$\Phi(t)$, supply rate and manipulate variable

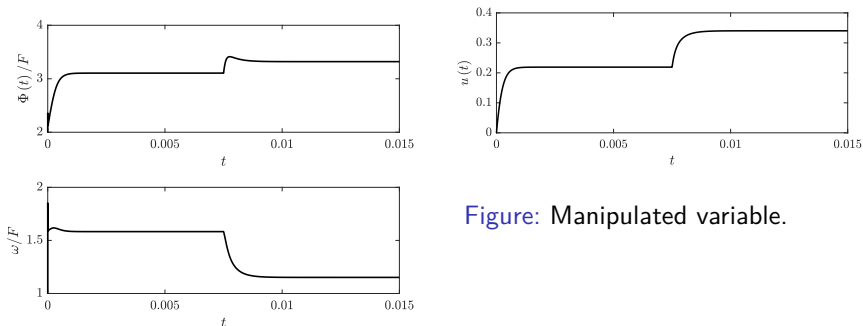


Figure: Manipulated variable.

Figure: $\Phi(t)$ function and supply rate per flow.



Space-time variation, entropy and internal entropy production

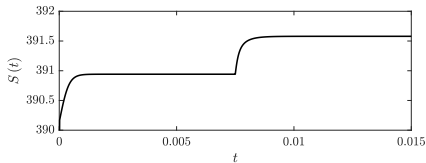


Figure: Entropy function.

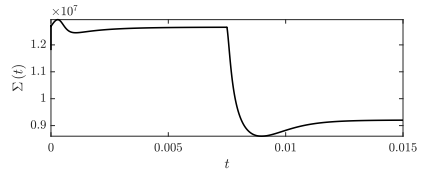


Figure: Internal entropy production.



Summary

- The proposed PI structure **smoothly tracks** a desired set point without offset.
- This structure is **robust** against uncertain terms, such as $\dot{\Sigma}$.
- Although the entropy function gives insight on dissipativity, it cannot give information on the stability of the desired set point.
- The internal entropy production gives the principles for further applications, such as optimization and its time derivative can determine the stability of a desired trajectory.
- Outlook
 - To define saturation constraints in order to overcome potential sign changes in the $\Phi(t)$ function.
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For Further Reading I

- García-Sandoval, J., Hudon, N., Dochain, D., 2016. Conservative and dissipative phenomena in thermodynamical systems stability. IFAC-PapersOnLine 49 (24), 28–33.
- Reilly, M., Schmitz, R., 1966. Dynamics of a tubular reactor with recycle: Part I. stability of the steady state. AIChE Journal 12 (1), 153–161.
- Willems, J. C., 1972. Dissipative dynamical systems part I: General theory. Archive for Rational Mechanics and Analysis 45 (5), 321–351.

