

Output Regulation for a class of linear ODE-Hyperbolic PDE-ODE systems

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Joint work with J. Redaud and F. Bribiesca-Argomedeo

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Why hyperbolic systems?

- **Conservation/balance** of scalar quantities when taking into account:
 - ▶ Evolution (e.g., **transport**) of conserved quantities in space and time
 - ▶ Finite **speed of propagation** (vs. heat equation)

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 - ▶ slow propagation speeds (e.g. traffic)
 - ▶ spatially dependent characteristics (e.g. composite materials)
 - ▶ anisotropic behavior (e.g. ferromagnetism)

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Mathematically, this may look something like:

$$\partial_t \rho(t, x) = \nabla f(t, x) + S(t, x), \quad \forall (t, x) \in [0, T] \times \Omega,$$

where ρ is the **quantity conserved**, f is a **flux density** and S is a **source term**.

Motivation

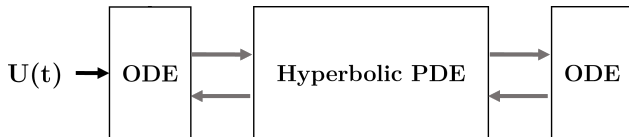
Many physical laws are **conservation/balance laws**, e.g. mass, charge, energy, momentum
[Bastin, Coron; 2016]



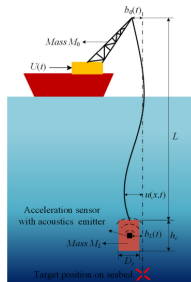
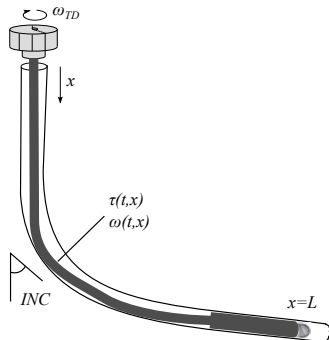
Why coupled and interconnected hyperbolic systems?

- Conservation/balance laws rarely appear isolated
 - ▶ Navier-Stokes → mass + energy + momentum
 - ▶ Propagation phenomena rarely occur in a single direction
- Systems modeled by hyperbolic PDEs do not exist in isolation, e.g.:
 - ▶ Electric transmission networks → interconnection of individual transmission lines
 - ▶ Mechanical vibrations in drilling devices → interconnection of different pipes
- Possible coupling with ODEs
 - ▶ actuator dynamics (e.g. pump, converter)
 - ▶ load dynamics (e.g. valve, motor)
 - ▶ sensor dynamics (e.g. flow-rate sensor, tachometer)

Examples of interconnected ODE-PDEs-ODE systems



Applications: drilling systems, deepwater construction vessels [Wang et al.]



- Interconnections of hyperbolic PDEs and ODEs are not a new problem.
- Many **constructive** control results based on the **backstepping approach**, e.g.:
 - ▶ Seminal paper [Krstic and Smyshlyayev, 2008]: re-interpretation of the classical Finite Spectrum Assignment [Manitius and Olbrot, 1979] (ODE + input delays)
 - ▶ Time-varying delays [Bekiaris-Liberis and Krstic, 2013, Bresch-Pietri, 2012],
 - ▶ Cascades of PDEs [Auriol et al., 2019]
 - ▶ Cascaded interconnections of hyperbolic PDE-ODE systems: [Aamo, 2012, Hasan et al., 2016, Zhou and Tang, 2012]

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 - ▶ Cascaded interconnections of hyperbolic PDE-ODE systems: [Aamo, 2012, Hasan et al., 2016, Zhou and Tang, 2012]
- For **fully-interconnected** (non-cascaded) systems some examples include:
 - ▶ stabilizing state-feedback control law in [Di Meglio et al., 2018, Wang et al., 2018]
 - ▶ output regulation for coupled linear wave-ODE systems [Deutscher and Gabriel, 2021]

- For ODE-hyperbolic PDE-ODE systems with **full interconnections** (non-cascade):
 - ▶ state feedback in [Bou Saba et al., 2017] for scalar PDE system (invertible input matrix)
 - ▶ output-feedback controller based on a Byrnes-Isidori normal form for the proximal ODE, as well as a relative degree one condition in [Deutscher et al., 2018]
 - ▶ strictly-proper state-feedback control law for scalar PDE in [Bou Saba et al., 2019] requiring minimum-phase assumption (not relative degree 1)
 - ▶ extended to output-feedback control for scalar PDE in [Wang and Krstic, 2020]
 - ▶ stabilizing observer-controller robust to delays in the case of a scalar proximal ODE in [Di Meglio et al., 2020]
- Some recent results have also been obtained for interconnected PDE systems with non-linear ODEs [Irscheid et al., 2021]

What you will see in this presentation

- **Output regulation** of a general class of ODE-PDE-ODE system
 - ▶ Finite-dimensional exo-system representing the reference trajectory and disturbance dynamics.
 - ▶ Backstepping approach: integral change of coordinates
 - ▶ Time delay representation and frequency analysis
 - ▶ Stabilizing control law in the absence of the disturbance

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- A **robustification procedure**
 - ▶ Low-pass filter to make the control law strictly proper
 - ▶ Frequency analysis

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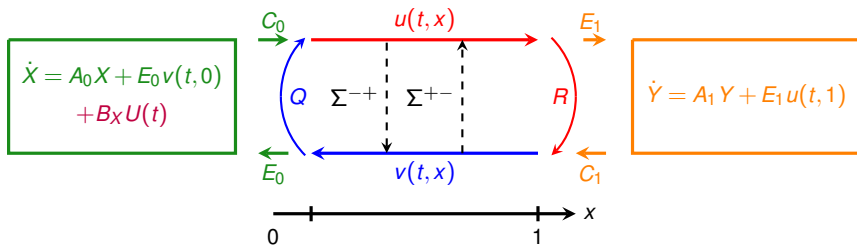
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- **Observer design**
 - ▶ Backstepping approach to simplify the dynamics
 - ▶ Luenberger-like observer with tuning operators
 - ▶ Frequency analysis
 - ▶ Output-feedback control law

System under consideration: ODE-PDE-ODE

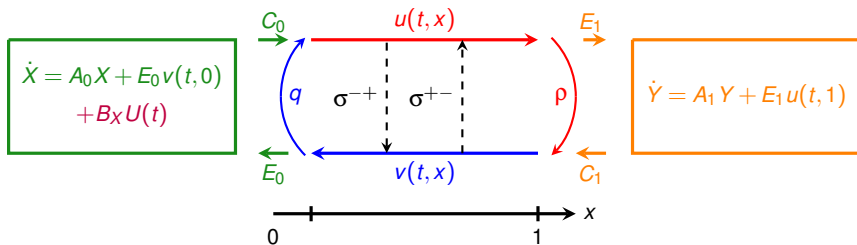
$$\left\{ \begin{array}{l} \dot{X}(t) = A_0 X(t) + E_0 v(t, 0) + B_X U(t), \\ \partial_t u(t, x) + \Lambda^+ \partial_x u(t, x) = \Sigma^{++}(x) u(t, x) + \Sigma^{+-}(x) v(t, x), \\ \partial_t v(t, x) - \Lambda^- \partial_x v(t, x) = \Sigma^{-+}(x) u(t, x) + \Sigma^{--}(x) v(t, x), \\ u(t, 0) = C_0 X(t) + Qv(t, 0), \quad v(t, 1) = Ru(t, 1) + C_1 Y(t), \\ \dot{Y}(t) = A_{11} Y(t) + E_1 u(t, 1), \end{array} \right.$$



- **Measurement:** $y(t) = C_{\text{mes}} Y(t)$
- Same concepts for scalar and non-scalar PDEs systems

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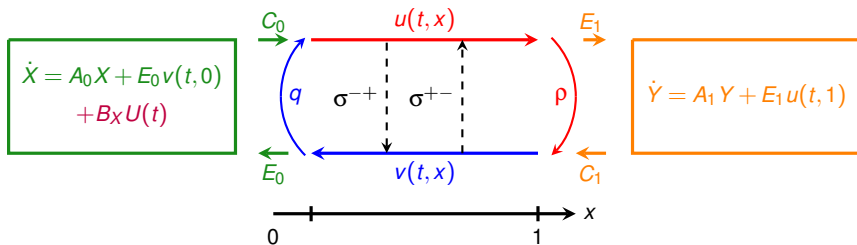
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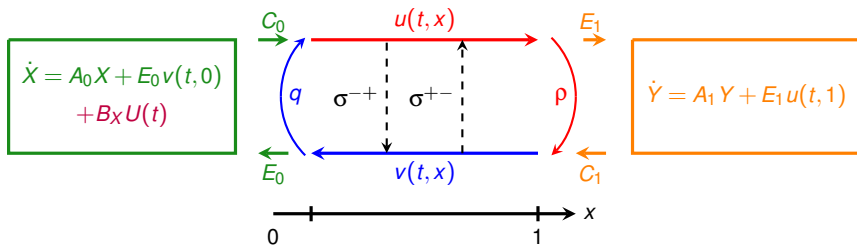
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- Initial conditions in H^1 with appropriate compatibility conditions \rightarrow **well-posedness**
- Stabilization in the sense of the L^2 -norm

System under consideration: well-posedness and stabilization objective

$$\left\{ \begin{array}{l} \dot{X}(t) = A_0 X(t) + E_0 v(t, 0) + B_X U(t), \\ \partial_t u(t, x) + \lambda \partial_x u(t, x) = \sigma^{+-}(x) u(t, x), \\ \partial_t v(t, x) - \mu \partial_x v(t, x) = \sigma^{-+}(x) u(t, x), \\ u(t, 0) = C_0 X(t) + qv(t, 0), \quad v(t, 1) = \rho u(t, 1) + C_1 Y(t), \\ \dot{Y}(t) = A_1 Y(t) + E_1 u(t, 1), \end{array} \right.$$

Well-posedness in open-loop

For every initial condition $(X_0, u_0, v_0, Y_0) \in \mathbb{R}^p \times H^1([0, 1], \mathbb{R}^2) \times \mathbb{R}^q$ that verifies the compatibility conditions

$$u_0(0) = C_0 X_0 + Qv_0(0), \quad v_0(1) = Ru_0(1) + C_1 Y_0$$

there exists one and one only (X, u, v, Y) which is a **solution to the open-loop Cauchy problem** (i.e., $U \equiv 0$).

Moreover, there exists $\kappa_0 > 0$ such that for every $(X_0, u_0, v_0, Y_0) \in \mathbb{R}^p \times H^1([0, 1], \mathbb{R}^2) \times \mathbb{R}^q$ satisfying the compatibility conditions, the unique solution verifies

$$\|(X(t), u(t, \cdot), v(t, \cdot), Y(t))\|_{\chi} \leq \kappa_0 e^{\kappa_0 t} \|(X_0, u_0, v_0, Y_0)\|_{\chi}, \quad \forall t \in [0, \infty).$$

where $\|(X(t), u(t, \cdot), v(t, \cdot), Y(t))\|_{\chi} = \sqrt{\|X(t)\|_{\mathbb{R}^p}^2 + \|u(t, \cdot)\|_{L^2}^2 + \|v(t, \cdot)\|_{L^2}^2 + \|Y(t)\|_{\mathbb{R}^q}^2}$.

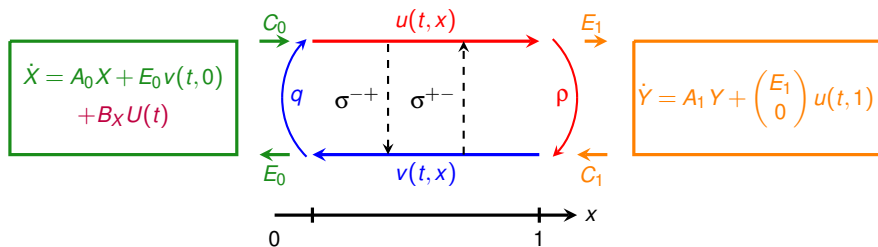
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Stabilization objective

Design a continuous control input that **exponentially stabilizes** the system in the sense of the L^2 -norm, i.e. there exist κ_0 and $\nu > 0$ such that for any initial condition $(X_0, u_0, v_0, Y_0) \in \mathbb{R}^p \times H^1([0, 1], \mathbb{R}^2) \times \mathbb{R}^q$, we have

$$\|(X(t), u(t, \cdot), v(t, \cdot), Y(t))\|_{\mathcal{X}} \leq \kappa_0 e^{-\nu t} \|(X_0, u_0, v_0, Y_0)\|_{\mathcal{X}}, \quad 0 \leq t$$

Output-regulation problem

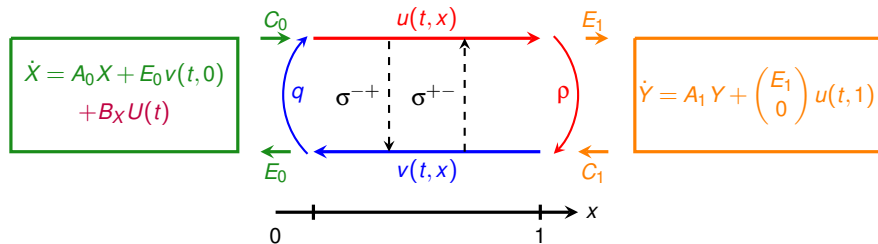


Augmented variable: $Y(t) = (Y_1^\top(t), Y_2^\top(t))^\top$

- Y_1 is the "real" ODE state
- Y_2 is an **exogenous input**: **disturbance** Y_{dist} and/or a **reference trajectory** Y_{ref}

$$\dot{Y}(t) = A_1 Y(t) + \begin{pmatrix} E_1 \\ 0_{q_2 \times 1} \end{pmatrix} u(t,1), \text{ with } A_1 = \begin{pmatrix} A_{11} & A_{12} \\ 0_{q_2 \times q_1} & A_{22} \end{pmatrix},$$

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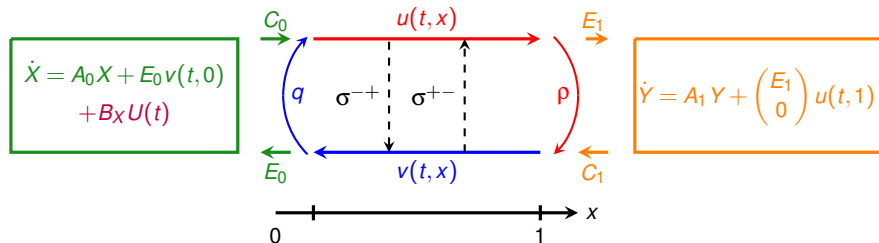
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Virtual output: $\varepsilon(t) = C_e Y(t) = (C_{e1} \quad C_{e2}) Y(t)$

Control objective

Design a control law $U(t)$ s.t. **the virtual output** $\varepsilon(t)$ exp. converges to zero.

Output-regulation problem



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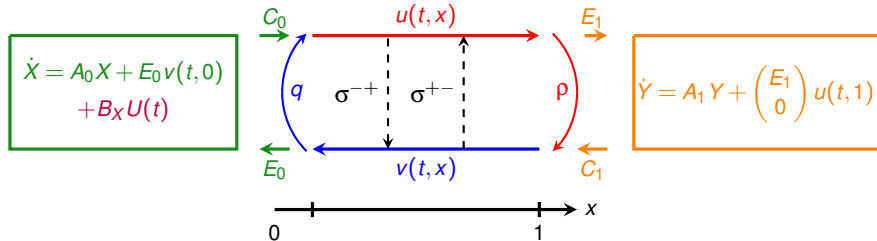
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- **Output regulation problem:** $C_{e1} \neq 0$, and $C_{e2} \equiv 0$: we want to regulate to zero a linear combination of components of $Y_1(t)$ in the presence of a disturbance $Y_2(t)$.

Output-regulation problem



Augmented variable: $Y(t) = (Y_1^\top(t), Y_2^\top(t))^\top$

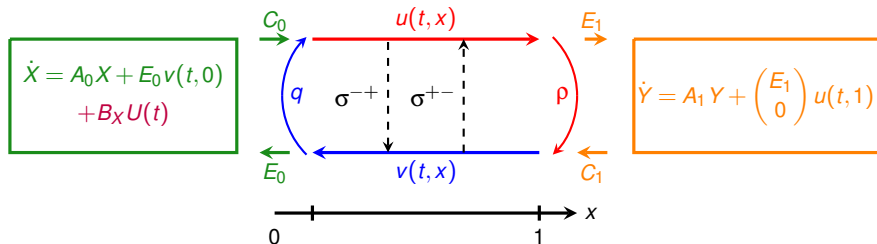
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- **Output tracking problem:** $C_{e1,j} - C_{e2,j} = 0$, (other components = 0): we want the j^{th} component of the output Y_1 to converge towards the j^{th} component of a known trajectory Y_2 .

Structural assumptions

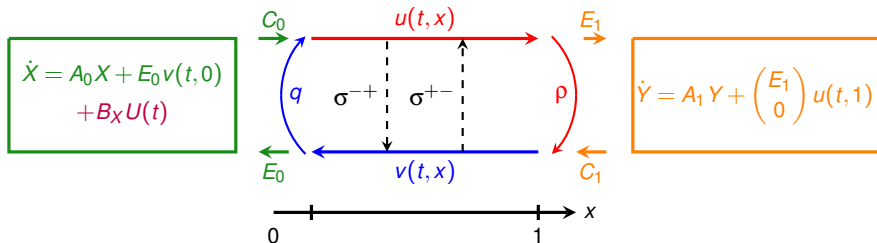


Assumption 1: Stabilizability

The pairs (A_0, B_0) and (A_{11}, E_1) are **stabilizable**, i.e. there exist $F_0 \in \mathbb{R}^{r \times p}$, $F_1 \in \mathbb{R}^{n \times q_1}$ such that $\bar{A}_0 \doteq A_0 + B_X F_0$ and $\bar{A}_{11} \doteq A_{11} + E_1 F_1$ are Hurwitz.

- Classical requirement found in most of the papers dealing with ODE-PDE-ODE
- Not overly conservative (necessary to stabilize Y , slightly conservative for X).

Structural assumptions



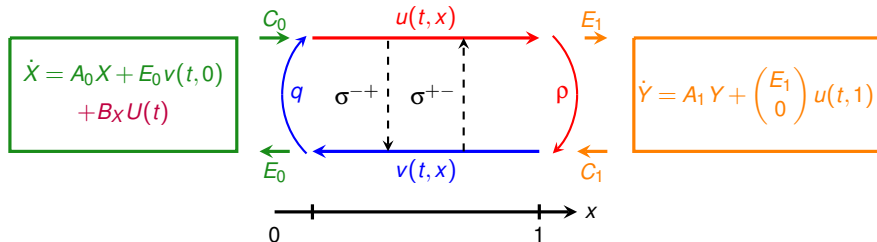
Assumption 2

For all $s \in \mathbb{C}_0$, the matrices (A_0, B_X, C_0) satisfy

$$\text{rank} \begin{pmatrix} \text{sld} - A_0 & B_X \\ C_0 & 0_{n \times r} \end{pmatrix} = p + 1 = p + n.$$

- The function $P_0(s) = C_0(\text{sld} - \bar{A}_0)^{-1} B_X$ does not have any zeros in \mathbb{C}^+
- **Stable right inverse** of $P_0(s)$

Structural assumptions

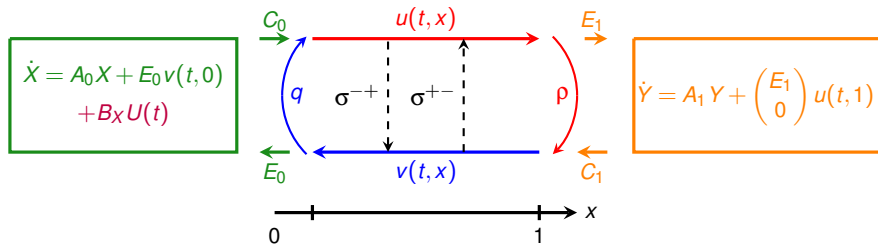


Assumption 3: Delay-robustness

The coefficients ρ and q verify $|\rho q| < 1$.

- No asymptotic chain of eigenvalues with non-negative real parts
- Necessary for (delay-) robust stabilization

Structural assumptions

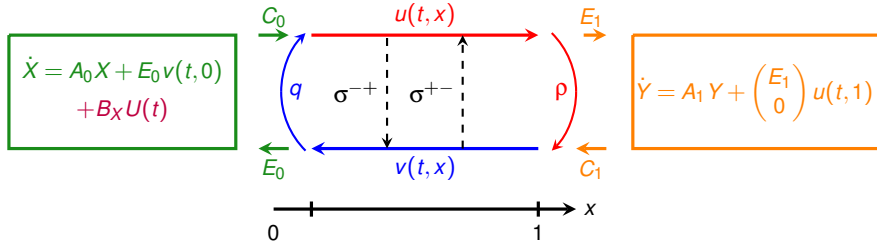


Assumption 4: detectability

The pairs (A_1, C) , (A_0, C_0) **are detectable** (i.e. there exist $L_0 \in \mathbb{R}^{p \times n}$ and $L_1 \in \mathbb{R}^{q \times d}$ such that $\tilde{A}_1 \doteq A_1 + L_1 C_{mes}$ and $\tilde{A}_0 \doteq A_0 + L_0 C_0$ are Hurwitz).

- Classical requirement found in most of the papers dealing with ODE-PDE-ODE
- Not overly conservative (necessary for reconstruction of X_0 , slightly conservative for Y).

Structural assumptions



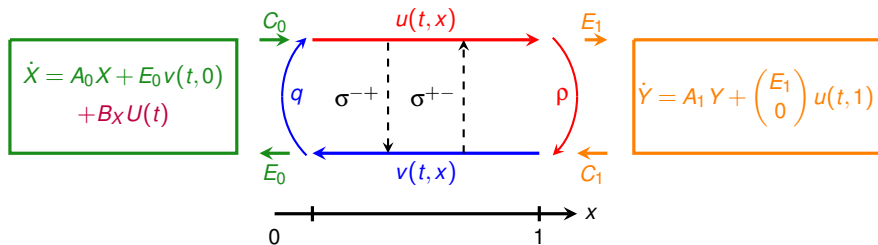
Assumption 5

For all $s \in \mathbb{C}^+$, the matrices (A_1, E_1, C) satisfy

$$\text{rank} \left(\begin{pmatrix} \text{sld} - A_1 & E_1 \\ C_{mes} & 0 \end{pmatrix} \right) = q + 1 = q + n. \quad (1)$$

- Necessary to independently reconstruct the different PDE boundary values by inverting the Y dynamics.
- The function $P_1(s) \doteq C_{mes}(\text{sld} - \tilde{A}_1)^{-1} E_1$ does not have any zeros in \mathbb{C}^+
- **Stable left-inverse** of $P_1(s)$

Structural assumptions



Assumption 6

The matrix A_{22} is marginally stable, i.e., all its eigenvalues have zero real parts. There exist matrices $T_a \in \mathbb{R}^{q_1 \times q_2}$, $F_a \in \mathbb{R}^{n \times q_2}$ solutions to the **regulator equations**:

$$\begin{cases} -A_{11} T_a + T_a A_{22} + A_{12} = -E_1 F_a, \\ -C_{e1} T_a + C_{e2} = 0. \end{cases}$$

- Non-resonance condition.
- A_{11} and A_{22} have disjoint spectra, and the number of outputs we regulate is coherent with the number of inputs.
- The matrices T_a, F_a can be computed using a Schur triangulation.

- Backstepping transformation to simplify the dynamics and the design of the control law.
- The regulation problem rewrites as a stabilization problem.
- Time-delay representation and frequency analysis.
- Low-pass filtering of the control law to make it **strictly proper**.

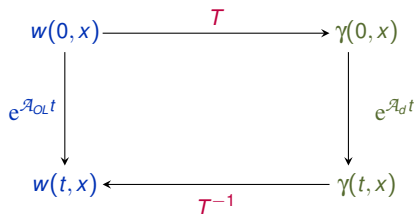
Backstepping methodology

- Map the original system to a *target system* for which the stability analysis is easier.
- Variable change: integral transformation, classically Volterra transform of the *second kind*

$$\alpha(t, x) = u(t, x) - \int_0^x K^{uu}(x, \xi)u(t, \xi) + K^{uv}(x, \xi)v(t, \xi)d\xi,$$

$$\beta(t, x) = v(t, x) - \int_0^x K^{vu}(x, \xi)u(t, \xi) + K^{vv}(x, \xi)v(t, \xi)d\xi,$$

Condensed form: $\gamma(t, x) = w(t, x) - \int_0^x K(x, y)w(t, y)dy.$



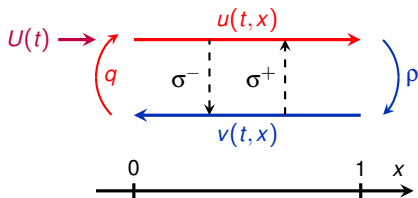
Limitations

- Choice of an adequate target system.
- Proof of existence and invertibility of an adequate backstepping transform.

Objective: Move the in-domain coupling terms at the actuated boundary.

$$u_t(t, x) + \lambda u_x(t, x) = \sigma^+ v(t, x),$$

$$v_t(t, x) - \mu v_x(t, x) = \sigma^- u(t, x).$$



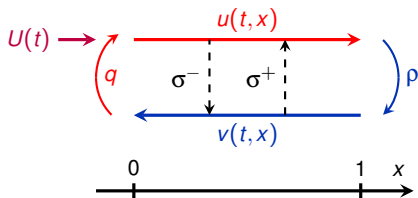
$$u(t, 0) = qv(t, 0) + U(t)$$

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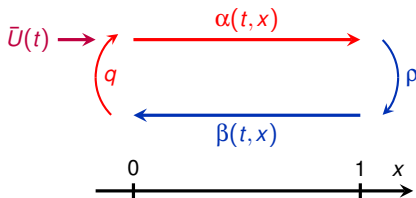


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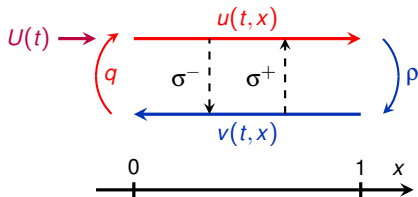
$$\alpha_t(t, x) + \lambda \alpha_x(t, x) = 0,$$

$$\beta_t(t, x) - \mu \beta_x(t, x) = 0.$$



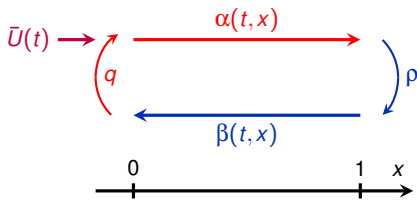
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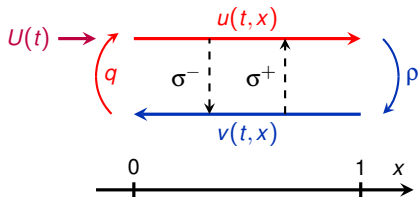


$$\alpha(t, 0) = q\beta(t, 0) + \bar{U}(t)$$
$$- \int_0^1 N^\alpha(\xi) \alpha(t, \xi) + N^\beta(\xi) \beta(t, \xi) d\xi.$$
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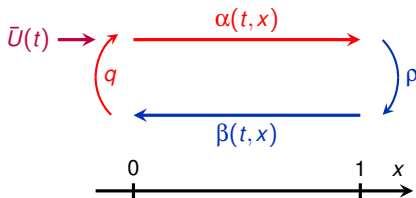


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$$\alpha_t(t, x) + \lambda \alpha_x(t, x) = 0,$$

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$$\alpha(t, 0) = q\beta(t, 0) + U(t)$$

$$- \int_0^1 N^\alpha(\xi) \alpha(t, \xi) + N^\beta(\xi) \beta(t, \xi) d\xi.$$

$$\beta(t, 1) = \rho \alpha(t, 1)$$

Natural control law

$$U(t) = -q\beta(t, 0) + \int_0^1 (N^\alpha(\xi) \alpha(t, \xi) + N^\beta(\xi) \beta(t, \xi)) d\xi.$$

$$X(t) = \xi(t) + \int_0^1 M^{12}(y)\alpha(t,y) + M^{13}(y)\beta(t,y)dy + [M^{14} \quad M^{15}] \eta(t),$$

$$u(t,x) = \alpha(t,x) + \int_x^1 M^{22}(x,y)\alpha(y) + M^{23}(x,y)\beta(y)dy + [M^{24}(x) \quad M^{25}(x)] \eta(t),$$

$$v(t,x) = \beta(t,x) + \int_x^1 M^{32}(x,y)\alpha(y) + M^{33}(x,y)\beta(y)dy + [M^{34}(x) \quad M^{35}(x)] \eta(t),$$

$$Y(t) = \eta(t).$$

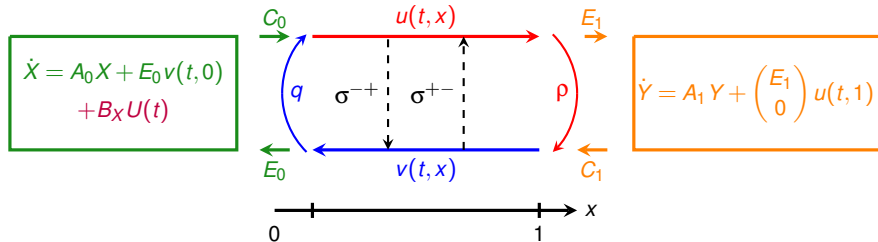
- Triangular transformation: invertible.

$$\begin{pmatrix} X(t) \\ u(t,x) \\ v(t,x) \\ Y(t) \end{pmatrix} = \begin{pmatrix} \mathbf{Id} & \int_0^1 M^{12}(y)dy & \int_0^1 M^{13}(y)dy & [M^{14} \quad M^{15}] \\ \mathbf{0} & \mathbf{Id} + \int_x^1 M^{22}(x,y)dy & \int_x^1 M^{23}(x,y)dy & [M^{24}(x) \quad M^{25}(x)] \\ \mathbf{0} & \int_x^1 M^{32}(x,y)dy & \mathbf{Id} + \int_x^1 M^{33}(x,y)dy & [M^{34}(x) \quad M^{35}(x)] \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & \mathbf{Id} \end{pmatrix} \begin{pmatrix} \xi(t) \\ \alpha(t,x) \\ \beta(t,x) \\ \eta(t) \end{pmatrix}$$

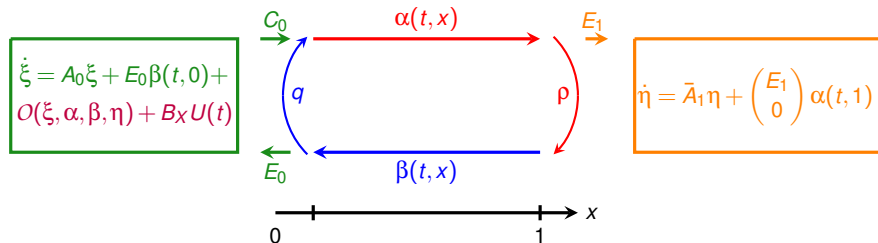
- Kernels are bounded functions.
- Unique solution due to the rank condition on C_0 .

Backstepping: Target system

Original system:



Target system:



Backstepping: Target system

Original system:

$$\left\{ \begin{array}{l} \dot{X}(t) = A_0 X(t) + E_0 v(t, 0) + B_X U(t), \\ \partial_t u(t, x) + \Lambda^+ \partial_x u(t, x) = \sigma^{+-}(x) u(t, x), \\ \partial_t v(t, x) - \Lambda^- \partial_x v(t, x) = \sigma^{-+}(x) u(t, x), \\ u(t, 0) = C_0 X(t) + qv(t, 0), \quad v(t, 1) = \rho u(t, 1) + C_1 Y(t), \\ \dot{Y}(t) = A_1 Y(t) + (E_1 \quad 0)^T u(t, 1), \end{array} \right.$$

Target system:

$$\left\{ \begin{array}{l} \dot{\xi}(t) = \bar{A}_0 \xi(t) + \bar{E}_1 \alpha(t, 1) + \bar{E}_0 \beta(t, 0) + M \eta(t) \\ \quad + \int_0^1 M_\alpha(y) \alpha(t, y) + M_\beta(y) \beta(t, y) dy + B_X \bar{U}(t), \\ \partial_t \alpha(t, x) + \Lambda^+ \partial_x \alpha(t, x) = 0, \\ \partial_t \beta(t, x) - \Lambda^- \partial_x \beta(t, x) = 0, \\ \alpha(t, 0) = C_0 \xi(t) + q \beta(t, 0), \quad \beta(t, 1) = \rho \alpha(t, 1), \\ \dot{\eta}(t) = \bar{A}_1 \eta(t) + (E_1 \quad 0)^T \alpha(t, 1), \end{array} \right.$$

$$\bar{A}_0 = A_0 + B_X F_0, \quad \bar{A}_1 = \begin{pmatrix} A_{11} + E_1 F_1 & A_{12} + E_1 (F_a + F_1 T_a) \\ 0 & A_{22} \end{pmatrix}$$

Advantages of the target system:

- Simplified in-domain couplings.
- Almost a "cascade structure"
- To stabilize the whole system, we can focus on the stabilization of ξ .

A cascade structure

$$\left\{ \begin{array}{l} \dot{\xi}(t) = \bar{A}_0 \xi(t) + \bar{E}_1 \alpha(t, 1) + \bar{E}_0 \beta(t, 0) + M \eta(t) \\ \quad + \int_0^1 M_\alpha(y) \alpha(t, y) + M_\beta(y) \beta(t, y) dy + B_x \bar{U}(t), \\ \partial_t \alpha(t, x) + \Lambda^+ \partial_x \alpha(t, x) = 0, \\ \partial_t \beta(t, x) - \Lambda^- \partial_x \beta(t, x) = 0, \\ \alpha(t, 0) = C_0 \xi(t) + q \beta(t, 0), \quad \beta(t, 1) = \rho \alpha(t, 1), \\ \dot{\eta}(t) = \bar{A}_1 \eta(t) + (E_1 \quad 0)^\top \alpha(t, 1), \end{array} \right.$$

Stability and regulation

If $C_0 \xi$ exp. converges to zero, then $\varepsilon(t) \rightarrow 0$. Furthermore, the trajectories are bounded.

A cascade structure

Assumption 6

The matrix A_{22} is marginally stable, i.e., all its eigenvalues have zero real parts. There exist matrices $T_a \in \mathbb{R}^{q_1 \times q_2}$, $F_a \in \mathbb{R}^{n \times q_2}$ solutions to the **regulator equations**:

$$\begin{cases} -A_{11}T_a + T_aA_{22} + A_{12} = -E_1F_a, \\ -C_{e1}T_a + C_{e2} = 0. \end{cases}$$

Stability and regulation

If ξ exp. converges to zero, then $\varepsilon(t) \rightarrow 0$. Furthermore, the trajectories are bounded.

Proof: If $C_0\xi$ converges to zero, then so does $\|(\alpha, \beta)\|_{L^2}$.

- We have

$$\begin{aligned} \dot{Y}_1 &= (A_{11} + E_1F_1)Y_1(t) + (A_{12} + E_1(F_a + F_1T_a))Y_2(t) + E_1\alpha(t, 1) \\ &= (A_{11} + E_1F_1)Y_1(t) + (A_{11}T_a - E_1F_a - T_aA_{22})Y_2(t) + E_1(F_a + F_1T_a)Y_2(t) + E_1\alpha(t, 1), \\ &\Rightarrow \underbrace{\dot{(Y_1 + T_aY_2)}}(t) = \bar{A}_{11}(Y_1 + T_aY_2) + \underbrace{E_1\alpha(t, 1)}_{\rightarrow 0}. \end{aligned}$$

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- $Y_1 + T_aY_2$ exp. stable $\Rightarrow C_e(Y_1 + T_aY_2)(t) = C_{e1}Y_1(t) + C_{e2}Y_2(t) = \varepsilon(t)$ goes to zero.

A cascade structure

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Stability and regulation

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- $Y_1 + T_aY_2$ exp. stable $\Rightarrow C_e(Y_1 + T_aY_2)(t) = C_{e1}Y_1(t) + C_{e2}Y_2(t) = \varepsilon(t)$ goes to zero.
- Invertibility + boundedness of the backstepping transf. implies boundedness of the state.

$$\alpha_t(t, x) + \lambda \alpha_x(t, x) = 0$$

$$\beta_t(t, x) - \mu \beta_x(t, x) = 0$$

$$\alpha(t, 0) = q\beta(t, 0) + C_0\xi(t)$$

$$\beta(t, 1) = \rho\alpha(t, 1)$$

$$\alpha_t(t, x) + \lambda \alpha_x(t, x) = 0 \rightarrow \text{Transport equation}$$

$$\beta_t(t, x) - \mu \beta_x(t, x) = 0 \rightarrow \text{Transport equation}$$

$$\alpha(t, 0) = q\beta(t, 0) + C_0\xi(t)$$

$$\beta(t, 1) = \rho\alpha(t, 1)$$

Time-delay representation

$$\alpha_t(t, x) + \lambda \alpha_x(t, x) = 0 \rightarrow \text{Transport equation}$$

$$\beta_t(t, x) - \mu \beta_x(t, x) = 0 \rightarrow \text{Transport equation}$$

$$\alpha(t, 0) = q\beta(t, 0) + C_0\xi(t)$$

$$\beta(t, 1) = \rho\alpha(t, 1)$$

Method of characteristics:

$$\alpha(t, x) = \alpha\left(t - \frac{x}{\lambda}, 0\right), \quad \beta(t, x) = \rho\alpha\left(t - \frac{(1-x)}{\mu} - \frac{1}{\lambda}, 0\right)$$

Difference Equation satisfied by $\alpha(t, 0)$

$$\alpha(t, 0) = \rho q \alpha(t - \tau, 0) + C_0 \xi(t), \quad t > \frac{1}{\lambda} + \frac{1}{\mu} = \tau$$

Using the Laplace transform: $(1 - \rho q e^{-\tau s})\alpha(s, 0) = C_0 \xi(s)$

We can kill the α and β terms to obtain ξ -terms!

Time-delay representation

$$\dot{\eta}(t) = \bar{A}_1 \eta(t) + (E_1 \quad 0)^\top \alpha(t, 1)$$

$$\dot{\xi}(t) = \bar{A}_0 \xi(t) + \bar{E}_1 \alpha(t, 1) + \bar{E}_0 \beta(t, 0) + M \eta(t) + \int_0^1 M_\alpha(y) \alpha(t, y) + M_\beta(y) \beta(t, y) dy + B_X \bar{U}(t).$$

Laplace transform on η_1

$$\eta_1(s) = (sI - \bar{A}_{11})^{-1} (\bar{A}_{12} \eta_2(s) + E_1 e^{-\frac{s}{\lambda}} \alpha(s, 0))$$

We can get rid of the η_1 -terms!

Time-delay representation

$$\dot{\eta}(t) = \bar{A}_1 \eta(t) + (E_1 \quad 0)^\top \alpha(t, 1)$$

$$\dot{\xi}(t) = \bar{A}_0 \xi(t) + \bar{E}_1 \alpha(t, 1) + \bar{E}_0 \beta(t, 0) + M \eta(t) + \int_0^1 M_\alpha(y) \alpha(t, y) + M_\beta(y) \beta(t, y) dy + B_X \bar{U}(t).$$

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We can get rid of the η_1 -terms!

Laplace transform on ξ

$$(sI - \bar{A}_0) \xi(s) = G(s) C_0 \xi(s) + H(s) \eta_2(s) + B_X \bar{U}(s),$$

$P_0 = C_0 (sI - \bar{A}_0)^{-1} B_X$ admits a stable right inverse P_0^+ .

$$C_0 \xi(s) = C_0 (sI - \bar{A}_0)^{-1} G(s) C_0 \xi(s) + C_0 (sI - \bar{A}_0)^{-1} H(s) \eta_2(s) + P_0(s) \bar{U}(s),$$

Time-delay representation

$$\dot{\eta}(t) = \bar{A}_1 \eta(t) + (E_1 \quad 0)^\top \alpha(t, 1)$$

$$\dot{\xi}(t) = \bar{A}_0 \xi(t) + \bar{E}_1 \alpha(t, 1) + \bar{E}_0 \beta(t, 0) + M \eta(t) + \int_0^1 M_\alpha(y) \alpha(t, y) + M_\beta(y) \beta(t, y) dy + B_X \bar{U}(t).$$

Laplace transform on η_1

$$\eta_1(s) = (\text{sld} - \bar{A}_{11})^{-1} (\bar{A}_{12} \eta_2(s) + E_1 e^{-\frac{s}{\lambda}} \alpha(s, 0))$$

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Stabilizing control law

$$\bar{U}(s) = \underbrace{-P_0^+(s) C_0(\text{sld} - \bar{A}_0)^{-1} G(s) C_0 \xi(s)}_{\text{stabilization}} - \underbrace{P_0^+(s) C_0(\text{sld} - \bar{A}_0)^{-1} H(s) \eta_2(s)}_{\text{disturbance rejection or tracking}}$$

Stabilizing control law

$$\begin{aligned}\bar{U}(s) &= \underbrace{-P_0^+(s)C_0(s\text{Id} - \bar{A}_0)^{-1}G(s)C_0\xi(s)}_{\text{stabilization}} - \underbrace{P_0^+(s)C_0(s\text{Id} - \bar{A}_0)^{-1}H(s)\eta_2(s)}_{\text{disturbance rejection or tracking}} \\ &= F_\xi(s)\xi(s) + F_\eta(s)\eta_2(s)\end{aligned}$$

- The control law may not be **strictly proper** due to $P_0^+(s) \rightarrow$ **Robustness issues**.
- We can make $F_\eta(s)$ strictly proper using our prior knowledge of the dynamics.
- We can make $F_\xi(s)$ strictly proper using a low-pass filter.

Filtering of the control input

$$F_{\xi}(s) = -P_0^+(s)C_0(sI - \bar{A}_0)^{-1}G(s)C_0, \quad F_{\eta}(s) = -P_0^+(s)C_0(sI - \bar{A}_0)^{-1}H(s)$$

Filtered control law

Let $w(s)$ be any low-pass filter, with a sufficiently high relative degree, and $0 < \delta < 1$ such that

$$\forall x \in \mathbb{R}, |1 - w(jx)| \leq \frac{1 - \delta}{\|G\|_{\infty} \bar{\sigma}(C_0(jxI - \bar{A}_0)^{-1})},$$

then $\bar{U}(s) = w(s)F_{\xi}(s)\xi(s) + \bar{F}_{\eta}(s)\eta_2(s)$ stabilizes $C_0\xi(s)$

Proof: Let $\Phi(s) = (1 - w(s))C_0(sI - \bar{A}_0)^{-1}G(s)$.

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- Φ is stable and strictly proper

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- Φ is stable and strictly proper
- $G(s)$ is unif. bounded, we have $\bar{\sigma}(G(jx)) \leq \|G\|_{\infty}$ for all x

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- Φ is stable and strictly proper
- $G(s)$ is unif. bounded, we have $\bar{\sigma}(G(jx)) \leq \|G\|_{\infty}$ for all x
- We have $\bar{\sigma}(\Phi(jx)) \leq 1 - \delta \Rightarrow \|\Phi\|_{\infty} < 1$

Filtering of the control input

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- We have $\bar{\sigma}(\Phi(jx)) \leq 1 - \delta \Rightarrow \|\Phi\|_{\infty} < 1$
- Characteristic equation $(1 - \Phi(s))C_0\xi(s) = 0 \rightarrow$ **exponential stability**

Filtering of the control input

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Filtered control law

Let $w(s)$ be any low-pass filter, with a sufficiently high relative degree, and $0 < \delta < 1$ such that

$$\forall x \in \mathbb{R}, |1 - w(jx)| \leq \frac{1 - \delta}{\|G\|_{\infty} \bar{\sigma}(C_0(jxI_d - \bar{A}_0)^{-1})},$$

then $\bar{U}(s) = w(s)F_{\xi}(s)\xi(s) + \bar{F}_{\eta}(s)\eta_2(s)$ stabilizes $C_0\xi(s)$

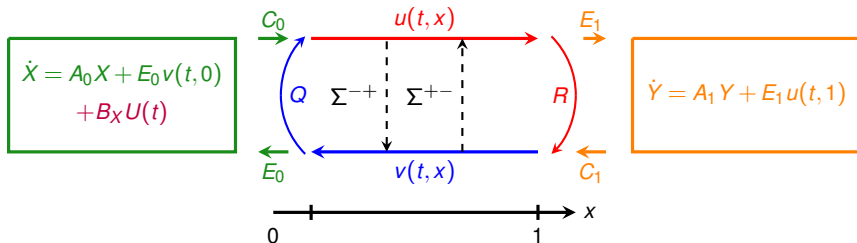
Proof: Let $\Phi(s) = (1 - w(s))C_0(sI_d - \bar{A}_0)^{-1}G(s)$.

- Φ is stable and strictly proper
- $G(s)$ is unif. bounded, we have $\bar{\sigma}(G(jx)) \leq \|G\|_{\infty}$ for all x
- We have $\bar{\sigma}(\Phi(jx)) \leq 1 - \delta \Rightarrow \|\Phi\|_{\infty} < 1$
- Characteristic equation $(1 - \Phi(s))C_0\xi(s) = 0 \rightarrow$ **exponential stability**

Strictly proper stabilizing control law!

- Backstepping transformation to simplify the dynamics and the design of the control law.
- The regulation problem rewrites as a stabilization problem.
- Time-delay representation and frequency analysis.
- Low-pass filtering of the control law to make it **strictly proper**.

$$\left\{ \begin{array}{l} \dot{X}(t) = A_0 X(t) + E_0 v(t, 0) + B_X U(t), \\ \partial_t u(t, x) + \Lambda^+ \partial_x u(t, x) = \Sigma^{++}(x) u(t, x) + \Sigma^{+-}(x) v(t, x), \\ \partial_t v(t, x) - \Lambda^- \partial_x v(t, x) = \Sigma^{-+}(x) u(t, x) + \Sigma^{--}(x) v(t, x), \\ u(t, 0) = C_0 X(t) + Qv(t, 0), \quad v(t, 1) = Ru(t, 1) + C_1 Y(t), \\ \dot{Y}(t) = A_{11} Y(t) + E_1 u(t, 1), \\ y = C_{mes} Y(t), \quad \dim(y) \geq \dim(u) \end{array} \right.$$



Problem statement

Design a **state observer** for the system based on the available measurement $y(t)$.

- Backstepping transformation to simplify the dynamics and the design of the observer.

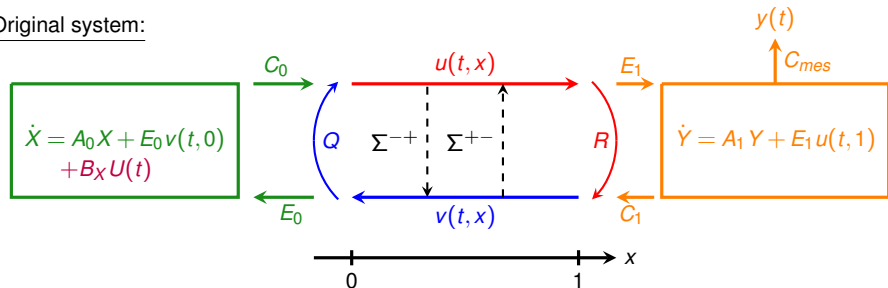
- Backstepping transformation to simplify the dynamics and the design of the observer.
- Luenberger-like observer with operators O_j that need to be tuned.

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- Design of the operators O_i to guarantee the exponential stability of the error system

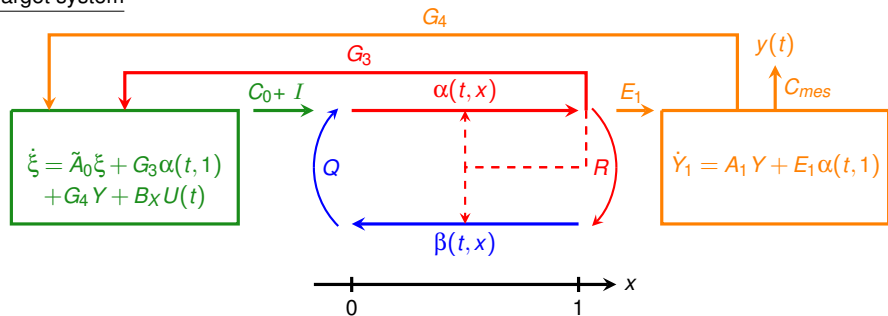
- Backstepping transformation to simplify the dynamics and the design of the observer.
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- Design of the operators O_i to guarantee the exponential stability of the error system
- Convergence of the observer state to the real state.

Backstepping: Target system

Original system:

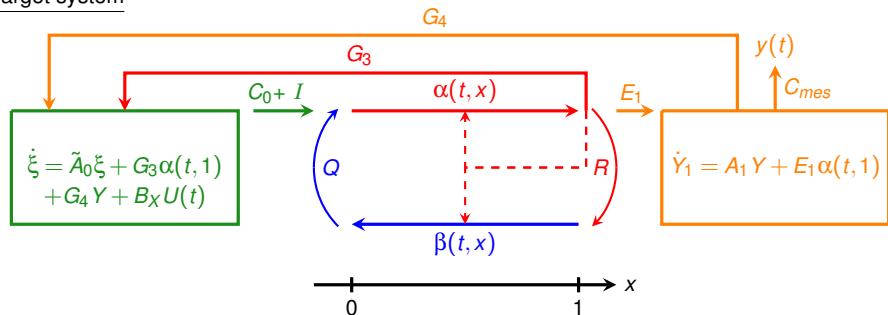


Target system



Backstepping: Target system

Target system



$$\dot{\xi}(t) = \tilde{A}_0 \xi(t) + G_3 \alpha(t, 1) + G_4 Y(t) + B_X U(t),$$

$$\alpha(t, 0) = Q\beta(t, 0) + C_0 \xi(t) + (Q\gamma_\beta(0) - \gamma_\alpha(0))Y(t) + \int_0^1 F^\alpha(y)\alpha(t, y) + F^\beta(y)\beta(t, y)dy,$$

$$\alpha_t(t, x) + \Lambda^+ \alpha_x(t, x) = G_1(x)\alpha(t, 1),$$

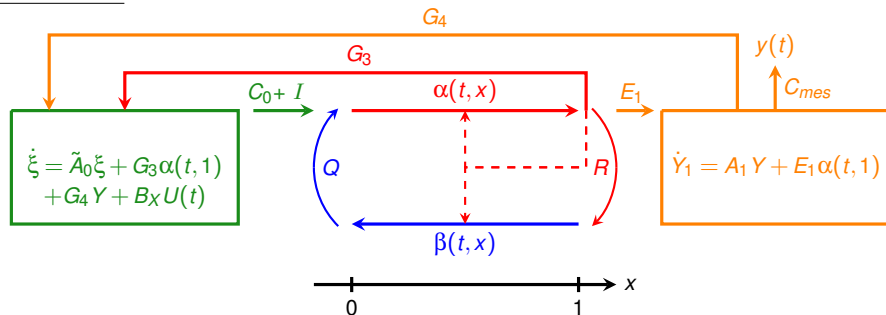
$$\beta_t(t, x) - \Lambda^- \beta_x(t, x) = G_2(x)\alpha(t, 1),$$

$$\beta(t, 1) = R\alpha(t, 1), \quad \dot{Y}(t) = A_1 Y(t) + E_1 \alpha(t, 1).$$

F^α strictly lower triangular

Backstepping: Target system

Target system



Advantages of the target system:

- Simplified in-domain couplings.
- Almost a "cascade structure" (except for the $\alpha(t, 1)$ -terms);
- Simplified observer design

$$X(t) = \xi(t) - \int_0^1 L_1(y)\alpha(y) + L_2(y)\beta(y)dy,$$

$$u(t, x) = \alpha(t, x) - \int_x^1 L^{\alpha\alpha}(x, y)\alpha(y)dy - \int_x^1 L^{\alpha\beta}(x, y)\beta(y)dy + \gamma_\alpha(x)Y(t),$$

$$v(t, x) = \beta(t, x) - \int_x^1 L^{\beta\alpha}(x, y)\alpha(y)dy - \int_x^1 L^{\beta\beta}(x, y)\beta(y)dy + \gamma_\beta(x)Y(t),$$

$$Y(t) = Y(t),$$

- Triangular transformation: invertible.
- Kernels are bounded functions.

Observer equations

System (ξ, α, β, Y)

$$\dot{\xi}(t) = \tilde{A}_0 \xi(t) + G_3 \alpha(t, 1) + G_4 Y(t) + B_X U(t),$$

$$\alpha(t, 0) = Q\beta(t, 0) + C_0 \xi(t) + (Q\gamma_\beta(0) - \gamma_\alpha(0))Y(t) + \int_0^1 F^\alpha(y)\alpha(t, y) + F^\beta(y)\beta(t, y)dy,$$

$$\alpha_t(t, x) + \Lambda^+ \alpha_x(t, x) = G_1(x)\alpha(t, 1),$$

$$\beta_t(t, x) - \Lambda^- \beta_x(t, x) = G_2(x)\alpha(t, 1),$$

$$\beta(t, 1) = R\alpha(t, 1), \quad \dot{Y}(t) = A_1 Y(t) + E_1 \alpha(t, 1).$$

System $(\hat{\xi}, \hat{\alpha}, \hat{\beta}, \hat{Y})$: O_i : **stable operators.**

$$\dot{\hat{\xi}}(t) = \tilde{A}_0 \hat{\xi}(t) + G_3 \hat{\alpha}(t, 1) + G_4 \hat{Y}(t) - O_0(\tilde{y}),$$

$$\hat{\alpha}(t, 0) = Q\hat{\beta}(t, 0) + C_0 \hat{\xi}(t) + (Q\gamma_\beta(0) - \gamma_\alpha(0))\hat{Y}(t)$$

$$+ \int_0^1 F^\alpha(y)\hat{\alpha}(t, y) + F^\beta(y)\hat{\beta}(t, y)dy - O_1(\tilde{y}),$$

$$\hat{\alpha}_t(t, x) + \Lambda^+ \hat{\alpha}_x(t, x) = G_1(x)\hat{\alpha}(t, 1) - O_\alpha(x, \tilde{y}),$$

$$\hat{\beta}_t(t, x) - \Lambda^- \hat{\beta}_x(t, x) = G_2(x)\hat{\alpha}(t, 1) - O_\beta(x, \tilde{y}),$$

$$\hat{\beta}(t, 1) = R\hat{\alpha}(t, 1), \quad \dot{\hat{Y}}(t) = A_1 \hat{Y}(t) + E_1 \hat{\alpha}(t, 1) - L_1 C \tilde{y},$$

$$\begin{aligned} \dot{\tilde{\xi}}(t) &= \tilde{A}_0 \tilde{\xi}(t) + G_3 \tilde{\alpha}(t, 1) + G_4 \tilde{Y}(t) + B_X U(t) O_0(\tilde{y}), \\ \tilde{\alpha}(t, 0) &= C_0 \tilde{\xi}(t) + Q \tilde{\beta}(t, 0) + (Q \gamma_\beta(0) - \gamma_\alpha(0)) \tilde{Y}(t) \\ &\quad + \int_0^1 F^\alpha(y) \tilde{\alpha}(t, y) + F^\beta(y) \tilde{\beta}(t, y) dy + O_1(\tilde{y}), \\ \tilde{\alpha}_t(t, x) + \Lambda^+ \tilde{\alpha}_x(t, x) &= G_1(x) \tilde{\alpha}(t, 1) + O_\alpha(x, \tilde{y}) \\ \tilde{\beta}_t(t, x) - \Lambda^- \tilde{\beta}_x(t, x) &= G_2(x) \tilde{\alpha}(t, 1) + O_\beta(x, \tilde{y}) \\ \tilde{\beta}(t, 1) &= R \tilde{\alpha}(t, 1), \quad \dot{\tilde{Y}}(t) = \tilde{A}_1 \tilde{Y}(t) + E_1 \tilde{\alpha}(t, 1). \end{aligned}$$

- **Objective:** Tune the gains O_i such that the error system exponentially converges to zero.

Lemma: Cascade structure of the error system

If $\tilde{\xi}(t)$, $\tilde{\alpha}(t, 1)$ and $\tilde{Y}(t)$ exponentially converge to zero, then the state $(\tilde{\xi}, \tilde{\alpha}, \tilde{\beta}, \tilde{Y})$ exponentially converges to zero. This implies the convergence of the observer state to the real state.

- Laplace transform of $\dot{Y}(t) = \tilde{A}_1 \tilde{Y}(t) + E_1 \tilde{\alpha}(t, 1)$:

$$(\text{sId} - \tilde{A}_1) \tilde{Y}(s) = E_1 \tilde{\alpha}(s, 1) \rightarrow \tilde{y}(s) = C_{mes}(\text{sId} - \tilde{A}_1)^{-1} E_1 \tilde{\alpha}(s, 1),$$

where \tilde{A}_1 is Hurwitz (**Assumption 4**) and $C_{mes}(\text{sId} - \tilde{A}_1)^{-1} E_1$ has no zeros in the RHP (**Assumption 2**)

Design of the operators O_i

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- $P_1(s) = C_{mes}(\text{sld} - \tilde{A}_1)^{-1} E_1$ has a stable left-inverse (**Assumption 4**):

$$\tilde{\alpha}(s, 1) = P_1^-(s) \tilde{y}(s), \quad \tilde{Y}(s) = (\text{sld} - \tilde{A}_1)^{-1} E_1 P_1^-(s) \tilde{y}(s)$$

Terms that are functions \tilde{Y} and $\tilde{\alpha}(s, 1)$ can be (exponentially) compensated using stable filters and values of $\tilde{y}(s)$.

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- We have $\dot{\tilde{\xi}}(t) = \tilde{A}_0 \tilde{\xi}(t) + G_3 \tilde{\alpha}(t, 1) + G_4 \tilde{Y}(t) + O_0(\tilde{y})$

$$O_0(\tilde{y}(s)) = -(G_3 P_1^-(s) + G_4 (\text{sld} - \tilde{A}_1)^{-1} E_1 P_1^-(s)) \tilde{y}(s) \Rightarrow (\text{sld} - \tilde{A}_0) \tilde{\xi}(s) = 0$$

Exponential convergence of $\tilde{\xi}$ to 0.

$$\alpha(s, 1) = P_1^-(s)\tilde{Y}(s), \quad \tilde{y}(s) = (\text{sld} - \tilde{A}_1)^{-1} E_1 P_1^-(s)\tilde{y}(s)$$

- We have $\tilde{\alpha}_t(t, x) + \Lambda^+ \tilde{\alpha}_x(t, x) = G_1(x)\tilde{\alpha}(t, 1) + O_\alpha(x, \tilde{y})$. Thus

$$O_\alpha(x, \tilde{y}) = -G_1(x)P_1^-(s)\tilde{y}(s) \Rightarrow \tilde{\alpha}_t(t, x) + \Lambda^+ \tilde{\alpha}_x(t, x) = 0 \Rightarrow \tilde{\alpha}_i(t, x) = \tilde{\alpha}_i\left(t - \frac{x}{\lambda_i}, 0\right).$$

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- We have $\tilde{\beta}_t(t, x) - \Lambda^- \tilde{\beta}_x(t, x) = G_2(x)\tilde{\alpha}(t, 1) + O_\beta(x, \tilde{y})$. Thus

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$$\Rightarrow \beta_j(t, x) = \sum_{k=1}^n R_{jk} \tilde{\alpha}_k\left(t - \frac{1-x}{\mu_j}, 1\right).$$

An Integral Difference Equation

- The function $\tilde{\alpha}(t, 0)$ verifies

$$\begin{aligned}\tilde{\alpha}_i(s, 0) &= ((Q\gamma_\beta(0) - \gamma_\alpha(0))\tilde{Y})_i + (O_1(\tilde{y}))_i + \sum_{k=1}^m \sum_{\ell=1}^n Q_{ik} R_{k\ell} e^{-\frac{s}{\mu_k} - \frac{s}{\lambda_\ell}} \tilde{\alpha}_\ell(s, 0) \\ &+ \int_0^1 \sum_{k=1}^m \sum_{\ell=1}^n F_{ik}^\beta(v) R_{k\ell} e^{-\frac{s(1-v)}{\mu_k}} \tilde{\alpha}_\ell(s, 1) dv \\ &+ \int_0^1 \sum_{j=1}^i F_{ij}^\alpha(v) \sum_{k=1}^m \sum_{\ell=1}^n Q_{jk} R_{k\ell} e^{-\frac{sv}{\lambda_j}} e^{-\frac{s}{\mu_k}} \tilde{\alpha}_\ell(s, 1) dv,\end{aligned}$$

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- Possible to recursively define $O_1(\tilde{y})$ such that

$$\tilde{\alpha}_i(t, 0) = \sum_{k=1}^m \sum_{\ell=1}^n Q_{ik} R_{k\ell} \tilde{\alpha}_\ell\left(t - \frac{1}{\mu_k} - \frac{1}{\lambda_\ell}, 0\right)$$

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- Exponential stabilization of $\tilde{\alpha}(t, 0)$ (and consequently of $\tilde{\alpha}(t, 1)$) due to **Assumption 3**.

Convergence of the observer

- The states $\tilde{\alpha}(t, 1)$ and $\tilde{\xi}$ exponentially converge to zero.
- We have $\dot{\tilde{Y}}(t) = \tilde{A}_1 \tilde{Y}(t) + E_1 \tilde{\alpha}(t, 1)$ with \tilde{A}_1 Hurwitz. Thus the state \tilde{Y} **exponentially converges to zero**.
- Stabilization of the error system.

Convergence of the observer

With the proposed operators $O_0, O_\alpha, O_\beta, O_1$, the observer state $(\hat{X}, \hat{u}, \hat{v}, \hat{Y}) = \mathcal{T}(\hat{\xi}, \hat{\alpha}, \hat{\beta}, \hat{Y})$ exponentially converges to (X, u, v, Y) , \mathcal{T} being the inverse backstepping transformation.

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- Possible to low-pass filter the measured output signal to use **strictly proper observer operators**
- The proposed observer could be combined with the previous state-feedback laws to obtain a strictly proper **output-feedback controller**.

Simulation results

Parameters:

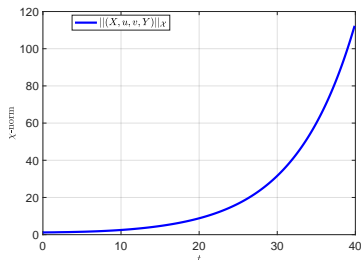
$$\lambda = 2, \mu = 0.7, \sigma^{+-} = 1, \sigma^{-+} = 0.5, \rho = 0.5, q = 1.2.$$

ODE dynamics in dimension $n = 4, m = 3, c = 2$

$$A_0 = \begin{bmatrix} 0 & 0.14 & 0 & 0.1 \\ 0 & 0 & 0.14 & 0 \\ 0.29 & -0.43 & 0.57 & 0.2 \\ 0 & 0 & 0 & -1.1 \end{bmatrix}, B_0 = \begin{bmatrix} 0 & 0 \\ 0 & -1 \\ 1 & -1 \\ 0 & 0 \end{bmatrix},$$

$$C_0 = \begin{bmatrix} 1 \\ 0 \\ 0 \\ -0.5 \end{bmatrix}^T, E_0 = \begin{bmatrix} 2 \\ -1 \\ 0.1 \\ 0 \end{bmatrix}, C_{11} = \begin{bmatrix} 0 \\ 1 \\ 0.5 \end{bmatrix}^T$$

$$A_{11} = \begin{bmatrix} 0.29 & 0.14 & 0 \\ 0.14 & 0 & 0.1 \\ 0 & 0 & -0.9 \end{bmatrix}, E_1 = \begin{bmatrix} -1 \\ 1 \\ 0 \end{bmatrix}.$$



Unstable system in open-loop.

We want to reject a sinusoidal disturbance

Simulation results

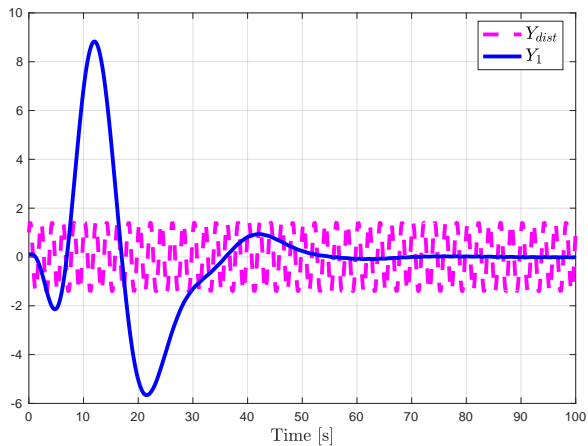


Figure: Evolution of the distal ODE state $Y_1(t)$ (blue) in the presence of a disturbance Y_{dist}

Simulation results

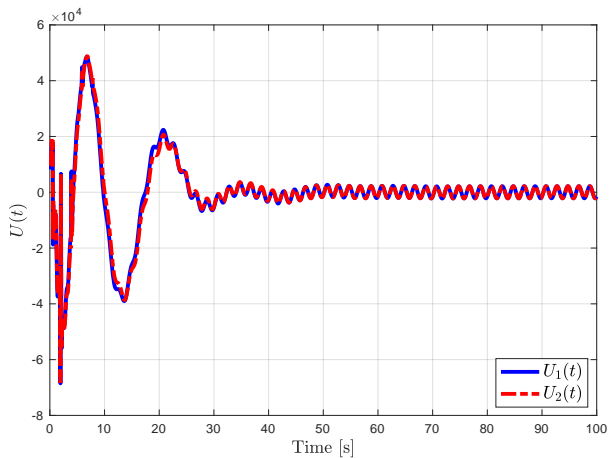


Figure: Evolution of the control inputs $U_1(t)$ (blue) and $U_2(t)$ (red)

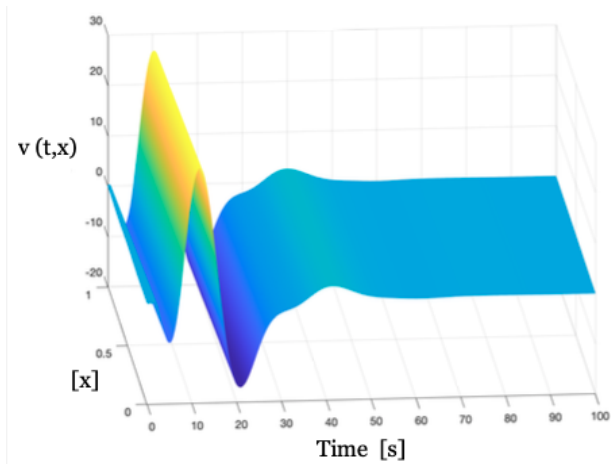


Figure: Evolution of the PDE state $v(t,x)$

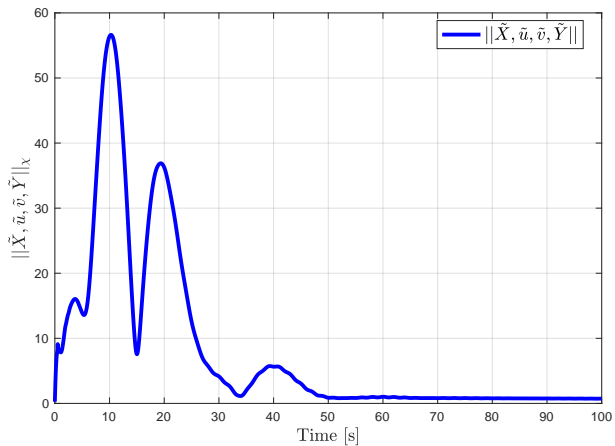


Figure: Evolution of the norm of the error state

- **Strictly proper dynamic** state-feedback controller for dist. rejection and trajectory tracking
 - ▶ Backstepping transformation to simplify the structure of the system
 - ▶ Frequency analysis to design the control law
 - ▶ Filtering techniques to guarantee robustness

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- **Luenberger-like observer** for the ODE-PDE-ODE system
 - ▶ Backstepping transformation and frequency analysis approach for the error system.
 - ▶ Output-feedback control law.
 - ▶ Computational effort?

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- **Perspectives?**
 - ▶ Model reduction?
 - ▶ Leverage the different assumptions?
 - ▶ Structure of the interconnection?

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