

Backstepping methods for boundary stabilization of 1-D hyperbolic balance laws

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Our hyperbolic balance laws is

$$\frac{\partial u}{\partial t} + A(u) \frac{\partial u}{\partial x} = F(u), \quad (t, x) \in [0, T] \times [0, L], \quad (1.1)$$

where,

- $u = (u_1, \dots, u_n)^T$ is a vector function of (t, x) ;
- $A(u)$ has n real eigenvalues $\lambda_i(u)$ ($i = 1, \dots, n$) and a complete set of left (resp. right) eigenvectors $l_i(u) = (l_{i1}(u), \dots, l_{in}(u))$ (resp. $r_i(u) = (r_{1i}(u), \dots, r_{ni}(u))^T$, ($i = 1, \dots, n$)):

$$l_i(u)A(u) = \lambda_i(u)l_i(u) \quad (\text{resp. } A(u)r_i(u) = \lambda_i(u)r_i(u).) \quad (1.2)$$

- $F(u) = (f_1(u), \dots, f_n(u))^T$ is a given vector function of u with $F(0) = 0$.

Remark

In general, we call the following systems

$$\frac{\partial u}{\partial t} + \frac{\partial g(u)}{\partial x} = F(u) \quad (1.3)$$

to be hyperbolic balance laws, where the flux $g := (g_1, \dots, g_n)$ is a vector function of u . Obviously, system (1.3) can be written in the quasilinear form as (1.1) with the Jacobian matrix

$$A(u) := D(g(u)). \quad (1.4)$$

Introduction

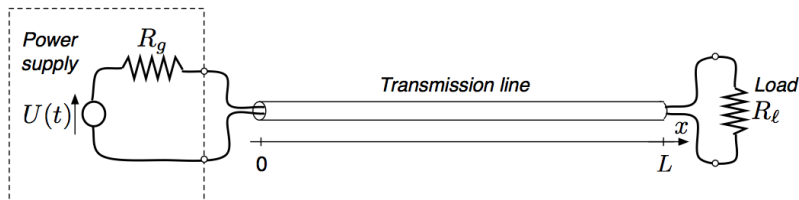
Many physical models are governed by linear and quasilinear hyperbolic balance laws, for example:

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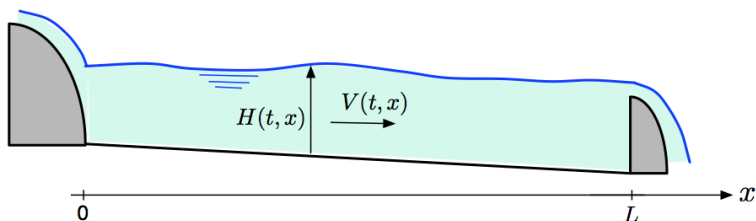
- The telegrapher equations (Heaviside, O. (1892))

$$\partial_t \begin{pmatrix} I \\ V \end{pmatrix} + \partial_x \begin{pmatrix} -L_e^{-1}V \\ -C_e^{-1}I \end{pmatrix} = - \begin{pmatrix} R_e L_e^{-1}I \\ G_e C_e^{-1}V \end{pmatrix}.$$



- The Saint-Venant equations (Barré de Saint-Venant (1871))

$$\frac{\partial H}{\partial t} + \frac{\partial}{\partial x}(HV) = 0,$$
$$\frac{\partial V}{\partial t} + \frac{\partial}{\partial x}\left(\frac{V^2}{2} + gH\right) = gS_b - CV^2H^{-1}.$$



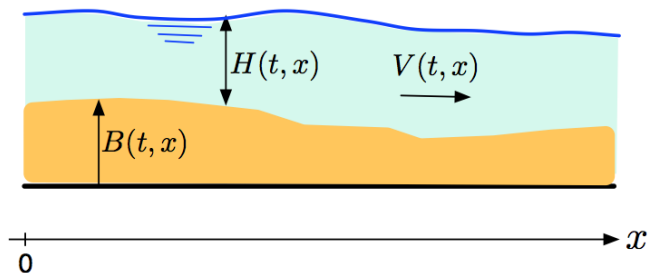
Introduction

- The Saint-Venant-Exner equations (Hudson-Sweby (2003))

$$\frac{\partial H}{\partial t} + V \frac{\partial H}{\partial x} + H \frac{\partial V}{\partial x} = 0,$$

$$\frac{\partial V}{\partial t} + V \frac{\partial V}{\partial x} + g \frac{\partial H}{\partial x} + g \frac{\partial B}{\partial x} = gS_b - C \frac{V^2}{H},$$

$$\frac{\partial B}{\partial t} + aV^2 \frac{\partial V}{\partial x} = 0.$$



- Heat exchangers (G.Bastin-J.-M.Coron (2016))

$$\partial_t H_1 + V_1 \partial_x H_1 + \frac{c^2}{g} \partial_x V_1 = 0$$

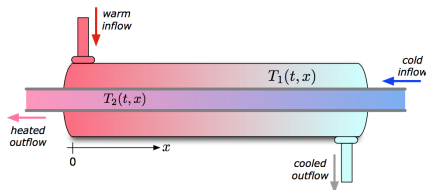
$$\partial_t V_1 + \partial_x (gH_1 + \frac{V_1^2}{2}) + \frac{C}{2d} V_1 |V_1| = 0$$

$$\partial_t T_1 + \partial_x (V_1 T_1) - k_1 (T_1 - T_2) - k_0 (T_1 - T_e) = 0$$

$$\partial_t H_2 + V_2 \partial_x H_2 + \frac{c^2}{g} \partial_x V_2 = 0,$$

$$\partial_t V_2 + \partial_x (gH_2 + \frac{V_2^2}{2}) + \frac{C}{2d} V_2 |V_2| = 0$$

$$\partial_t T_2 + \partial_x (V_2 T_2) + k_2 (T_1 - T_2) = 0$$



All these models can be rewritten as inhomogeneous hyperbolic systems:

$$\frac{\partial u}{\partial t} + A \frac{\partial u}{\partial x} = Bu \text{ (Telegrapher equations)}$$

$$\frac{\partial u}{\partial t} + A(u) \frac{\partial u}{\partial x} = F(u) \text{ (Saint-Venant(-Exner), Heat exchangers equations)}$$

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Remarks

- u is a vector;

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Remarks

- u is a vector;
- Balance Laws : $B \neq 0$, $F(u) \neq 0$, otherwise conservation laws.

Assumption

- Suppose that there is no zero eigenvalues for the matrix A .

Problem (Boundary Stabilization)

How to find a boundary feedback law such that the solution $u = u(t, x)$ of the hyperbolic systems with any given initial data satisfies

$$u(t, \cdot) \rightarrow 0, \quad \text{as } t \rightarrow +\infty? \quad (1.5)$$

Remarks

- Exponential stability: $\exists C, \lambda > 0$, such that

$$\|u(t, \cdot)\|_X \leq C e^{-\lambda t} \|u(0, \cdot)\|_X, \quad \forall t > 0 \quad (1.6)$$

- Finite-time stability: $\exists t_F$, such that

$$u(t, \cdot) \equiv 0, \quad \forall t \geq t_F. \quad (1.7)$$

Homogeneous case: (i.e. $B \equiv 0$, $F(u) \equiv 0$)

(1) **Characteristic method (T.T. Li, T.H. Qin (1983, 1985, 1994))**

- Quasilinear hyperbolic systems

$$u_t + A(u)u_x = 0$$

with the boundary conditions

$$\begin{pmatrix} u_-(t, 1) \\ u_+(t, 0) \end{pmatrix} = \mathcal{F} \begin{pmatrix} u_-(t, 0) \\ u_+(t, 1) \end{pmatrix} \quad (1.8)$$

- Framework of solution: C^1 norm,
- Local exponential stability (i.e. $\|u(0, \cdot)\|_{C^1}$ is suitably small);
- Boundary is "dissipative":

$$\rho_\infty(\mathcal{F}'(0)) < 1 \quad (1.9)$$

$$\mathcal{F}(0) = 0 \quad \text{and} \quad \rho_\infty(\mathcal{F}'(0)) := \inf\{\|\Delta \mathcal{F}'(0) \Delta^{-1}\|_\infty; \Delta \in \mathcal{D}_{n,+}\},$$

where $\mathcal{D}_{n,+}$ denotes the set of $n \times n$ real diagonal matrices with strictly positive diagonal elements.

Homogeneous case: (i.e. $B \equiv 0$, $F(u) \equiv 0$)

(2) **Control Lyapunov Functions method (G.Bastin, J.-M.Coron, B. d'Andréa-Noel (1999, 2007, 2008, 2014))**

- Quasilinear hyperbolic systems,

$$u_t + A(u)u_x = 0 \quad (1.10)$$

with the boundary conditions

$$\begin{pmatrix} u_-(t, 1) \\ u_+(t, 0) \end{pmatrix} = \mathcal{F} \begin{pmatrix} u_-(t, 0) \\ u_+(t, 1) \end{pmatrix} \quad (1.11)$$

- Local exponential stability.
- Boundary is "dissipative":

$$\rho_\infty(\mathcal{F}'(0)) < 1 \text{ for } C^1 \text{ norm;} \quad (1.12)$$

$$\rho_2(\mathcal{F}'(0)) < 1 \text{ for } H^2 \text{ norm.} \quad (1.13)$$

Complements for hyperbolic balance laws

- Characteristic method and Control Lyapunov Functions method : hyperbolic balance laws,

$$u_t + A(u)u_x = F(u) \quad (1.14)$$

can be exponentially stabilized by boundary feedback **provided** $\|\nabla F(0)\|$ **is small enough** (see T.T. Li (1994), J.-M. Coron-G. Bastin-B. d'Andréa-Novel (2008) and C. Prieur *et.al.* (2008)).

Complements for hyperbolic balance laws

What happens if $\|\nabla F(0)\|$ is not small?

- Characteristic method: hyperbolic balance laws

$$u_t + A(u)u_x = F(u) \quad (1.15)$$

exponentially decays to zero if both **boundary conditions** and $F(u)$ are “dissipative” in some sense (see C.M. Liu and Y.Z. Li (2015)).

- Control Lyapunov Functions method : A sufficient and necessary condition is given for the existence of basic quadratic strict control Lyapunov function for 2×2 linear hyperbolic balance laws. (see G.Bastin-J.M.Coron (2011))

Introduction

All the above works are based on the **static boundary output feedback** (i.e a feedback of the state values at the boundaries only).

However, **static boundary output feedback** can not treat all inhomogeneous case.

Counter example (G.Bastin-J.-M.Coron, 2016)

If

$$L \geq \frac{\pi}{c}. \quad (1.16)$$

there is no $k \in \mathbb{R}$ such that the equilibrium $(0, 0)^T \in L^2(0, L)^2$ is exponentially stable for the closed loop system

$$\begin{aligned} \partial_t S_1 + \partial_x S_1 + cS_2 &= 0, \\ \partial_t S_2 - \partial_x S_2 + cS_1 &= 0, \quad t \in [0, +\infty), \quad x \in [0, 1], \\ S_1(t, 0) &= kS_2(t, 0), \quad S_2(t, L) = S_1(t, L). \end{aligned} \quad (1.17)$$

Inhomogeneous case: (i.e. $B \neq 0$ and $F(u) \neq 0$)

- **Backstepping Method (J.-M. Coron-R. Vazquez-M. Krstic-G. Bastin (2011, 2013))**
 - 2×2 linear hyperbolic balance laws;

$$u_t + Au_x = Bu \quad (1.18)$$

- $B \in \mathcal{M}_{2,2}$ and $A = \text{diag}(-\lambda_1, \lambda_2)$ with $\lambda_1, \lambda_2 > 0$;
- Full-state feedback

$$u_-(t, L) = \int_0^L k(L, \xi)u(t, \xi)d\xi \quad (1.19)$$

- Finite-time stability.

Remark

The backstepping method can be extended to deal with the boundary stabilization problem of **inhomogeneous quasilinear** 2×2 hyperbolic systems (See J.-M. Coron-R. Vazquez-M. Krstic-G. Bastin (2013) and R. Vazquez-J.-M. Coron-M. Krstic-G. Bastin (2011))

Inhomogeneous case: ($n \times n$ hyperbolic balance laws)

- **Backstepping Method (F. Di Meglio-R.Vazquez-M.Krstic, 2013)**
 - 1 of the PDEs is controlled at its boundary and $n - 1$ other PDEs, which convect in the opposite direction, are not controlled and all have arbitrary interconnections, e.g.

$$u_t + Au_x = Bu \quad (1.20)$$

where $A = \text{diag}(-\lambda_1, -\lambda_2, \dots, -\lambda_{n-1}, \lambda_n)$ with

$$\lambda_i > 0, \quad i = 1, \dots, n.$$

and the only boundary feedback control is acting on $u_n(t, 0)$.

Unfortunately, the method presented in J.-M. Coron *et.al.* (2013) and R. Vazquez *et.al.* (2011,2013) and Di Meglio *et.al.* (2011, 2013) can not be directly extended to $n \times n$ systems, especially when several states convecting in the same direction are controlled.

Open Question (J.-M.Coron-R. Vazquez-M. Krstic-G.Bastin, SICON, 2013)

Can we stabilize the general inhomogeneous hyperbolic systems by multi-boundary feedback controls?

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We consider the following general linear hyperbolic system

$$u_t(t, x) + \Lambda^+ u_x(t, x) = \Sigma^{++} u(t, x) + \Sigma^{+-} v(t, x) \quad (2.1)$$

$$v_t(t, x) - \Lambda^- v_x(t, x) = \Sigma^{-+} u(t, x) + \Sigma^{--} v(t, x) \quad (2.2)$$

where $u = (u_1 \cdots u_n)^T$, $v = (v_1 \cdots v_m)^T$. and

$$\Lambda^+ = \text{diag}(\lambda_1 \cdots \lambda_n) \quad \Lambda^- = \text{diag}(\mu_1 \cdots \mu_m) \quad (2.3)$$

with

$$-\mu_1 < \cdots < -\mu_m < 0 < \lambda_1 \leq \cdots \leq \lambda_n \quad (2.4)$$

$\Sigma^{\pm\pm}$ are matrices and without loss of generality, we assume that we assume that

$$\forall j = 1, \dots, m \quad \sigma_{jj}^{\bar{-}\bar{-}} = 0, \quad (2.5)$$

Remark

One can do some coordinate transformation of v in order to guarantee (2.5).

The boundary conditions are as follows

$$u(t, 0) = Q_0 v(t, 0), \quad v(t, 1) = R_1 u(t, 1) + U(t) \quad (2.6)$$

where Q_0 and R_1 are constant matrices. $U(t) = (U_1, \dots, U_m)^T$ are boundary controls.

Goal

Our objective is to design a feedback control law for $U(t)$ in order to ensure that the closed-loop system vanishes in finite time.

What is Backstepping method?

- Mapping the original system to a target system which has “good” property (e.g. finite-time stable or exponential stable) by using an invertible backstepping transformation.

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- Mapping the original system to a target system which has “good” property (e.g. finite-time stable or exponential stable) by using a invertible backstepping transformation.

Difficulty of this method

- How to choose a good transformation?
- How to choose a suitable target system?

Review of 2×2 (i.e. $\Lambda \in \mathcal{M}_{2,2}$) case (JMC *et.al.* (2013) and R. Vazquez *et.al.* (2011))

- Transformation:

$$u(t, x) = \gamma(t, x) - \int_0^x K(x, \xi) \gamma(t, \xi) d\xi, \quad (2.7)$$

- Target system

$$\gamma_t(t, x) + \Lambda \gamma_x(t, x) = 0. \quad (2.8)$$

- The K-kernel is a matrix function of C^2 on the domain

$$\mathcal{T} = \{(x, \xi) | 0 \leq \xi \leq x \leq 1\}. \quad (2.9)$$

Unfortunately, we can not deal with the general $n \times n$ cases by using the above transformation and target system. If so, an overdetermined problem appears to the K-kernel.

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Question

- Is it possible to change only the transformation or the target system to achieve our purpose?

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Question

- Is it possible to change only the transformation or the target system to achieve our purpose?

Answer: Yes!

New target system

$$\alpha_t(t, x) + \Lambda^+ \alpha_x(t, x) = \Sigma^{++} \alpha(t, x) + \Sigma^{+-} \beta(t, x) + \int_0^x C^+(x, \xi) \alpha(\xi) d\xi + \int_0^x C^-(x, \xi) \beta(\xi) d\xi \quad (2.10)$$

$$\beta_t(t, x) - \Lambda^- \beta_x(t, x) = G(x) \beta(0) \quad (2.11)$$

with the following boundary conditions

$$\alpha(t, 0) = Q_0 \beta(t, 0), \quad \beta(t, 1) = 0 \quad (2.12)$$

where C^+ and C^- are L^∞ matrix functions on the domain

$$\mathcal{T} = \{0 \leq \xi \leq x \leq 1\}, \quad (2.13)$$

while $G \in L^\infty(0,1)$ is a lower triangular matrix with the following structures

$$G(x) = \begin{pmatrix} 0 & \cdots & \cdots & 0 \\ g_{2,1}(x) & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \vdots \\ g_{m,1}(x) & \cdots & g_{m,m-1}(x) & 0 \end{pmatrix}. \quad (2.14)$$

Finite-time stabilization (LH, F. Di Meglio, R. Vazquez and M.Krstic (2015 a))

The zero equilibrium of the target system is reached in finite time $t = t_F$, where

$$t_F := \frac{1}{\lambda_1} + \sum_{j=1}^m \frac{1}{\mu_j}. \quad (2.15)$$

Sketch of the proof

It is easy to see that β vanishes in finite time t_1 with

$$t_1 = \sum_{r=1}^m \frac{1}{|\lambda_r(s)|} ds. \quad (2.16)$$

From the time $t = t_1$ on, we find α becomes the solution of the following system

$$\alpha_t(t, x) + \Lambda^+ \alpha_x(t, x) = \Sigma^{++} \alpha(t, x) + \int_0^x C^+(x, \xi) \alpha(\xi) d\xi \quad (2.17)$$

with the boundary conditions

$$\alpha(t, 0) = 0. \quad (2.18)$$

Changing the status of t and x , and Equations (2.17) can be rewritten as

$$\alpha_x(t, x) + (\Lambda^+)^{-1} \alpha_t(t, x) = (\Lambda^+)^{-1} \Sigma^{++} \alpha(t, x) + \int_0^x (\Lambda^+)^{-1} C^+(x, \xi) \alpha(\xi) d\xi \quad (2.19)$$

with the initial condition (2.18).

Sketch of the proof

Then by the uniqueness of the system (2.18),(2.19), and noting the order of the transport speeds of the α -system (see (2.4)), this yields that α identically vanishes for

$$t \geq \frac{1}{\lambda_1} + \sum_{j=1}^m \frac{1}{\mu_j} \quad (2.20)$$

Open Question

How to reduce the finite-time-control time t_F ?

Backstepping transformation

We consider the following backstepping (Volterra) transformation

$$\alpha(t, x) = u(t, x) \tag{2.21}$$

$$\beta(t, x) = v(t, x) - \int_0^x [K(x, \xi)u(\xi) + L(x, \xi)v(\xi)] d\xi \tag{2.22}$$

where the kernels to be determined K and L are defined on the triangular domain \mathcal{T} .

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where the kernels to be determined K and L are defined on the triangular domain \mathcal{T} .

Important: $\alpha(t, 0) = u(t, 0), \beta(t, 0) = v(t, 0)$

Backstepping transformation

The original system (2.1) is mapped into the target system (2.6) if K and L satisfy the following equations

for $1 \leq i \leq m, 1 \leq j \leq n$

$$\mu_i \partial_x K_{ij}(x, \xi) - \lambda_j \partial_\xi K_{ij}(x, \xi) = \sum_{k=1}^n \sigma_{kj}^{++} K_{ik}(x, \xi) + \sum_{p=1}^m \sigma_{pj}^{-+} L_{ip}(x, \xi) \quad (2.23)$$

for $1 \leq i \leq m, 1 \leq j \leq m$

$$\mu_i \partial_x L_{ij}(x, \xi) + \mu_j \partial_\xi L_{ij}(x, \xi) = \sum_{p=1}^m \sigma_{pj}^{--} L_{ip}(x, \xi) + \sum_{k=1}^n \sigma_{kj}^{+-} K_{ik}(x, \xi) \quad (2.24)$$

along with the following set of boundary conditions

$$K_{ij}(x, x) = -\frac{\sigma_{ij}^{-+}}{\mu_i + \lambda_j} \triangleq k_{ij} \quad \text{for } 1 \leq i \leq m, \quad 1 \leq j \leq n \quad (2.25)$$

$$L_{ij}(x, x) = -\frac{\sigma_{ij}^{--}}{\mu_i - \mu_j} \triangleq l_{ij} \quad \text{for } 1 \leq i, j \leq m, \quad i \neq j \quad (2.26)$$

$$\mu_j L_{ij}(x, 0) = \sum_{k=1}^n \lambda_k K_{ik}(x, 0) q_{k,j} \quad \text{for } 1 \leq i \leq j \leq m. \quad (2.27)$$

Backstepping transformation

To ensure well-posedness of the kernel equations, we add the following artificial boundary conditions for $L_{ij}(i > j)$

$$L_{ij}(1, \xi) = l_{ij}, \text{ for } 1 \leq j < i \leq m. \quad (2.28)$$

Remark

We can select that

$$L_{ij}(1, \xi) = 0, \text{ for } 1 \leq j < i \leq m. \quad (2.29)$$

The existence of K and L

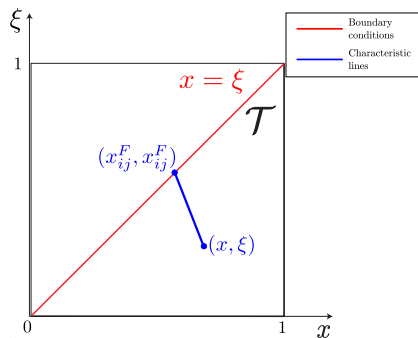
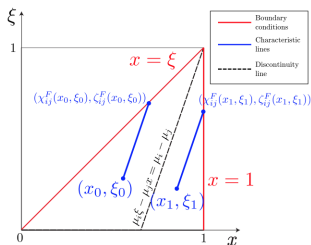
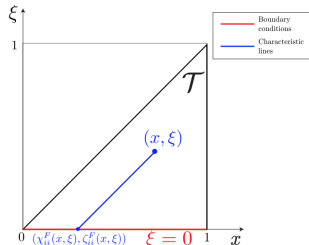


Figure 1 : Characteristic lines of the K kernels

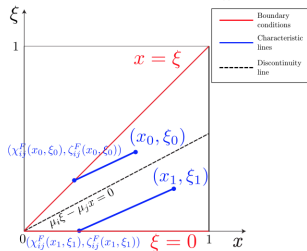
The existence of K and L



(a) Characteristic lines of the kernels L_{ij} for $i > j$



(b) Characteristic lines of the kernels L_{ii}



(c) Characteristic lines of the kernels L_{ij} for $i < j$

The existence of K and L

Successive approximation method (LH, F. DiMeglio, R. Vazquez and M. Krstic (2015 a))

$$K(x, \xi) = \sum_{q=0}^{+\infty} \Delta \mathbf{K}^q(x, \xi) \quad (2.30)$$

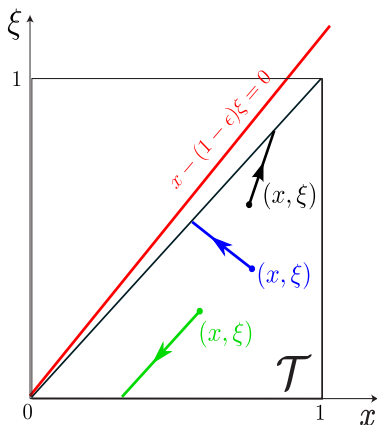
$$L(x, \xi) = \sum_{q=0}^{+\infty} \Delta \mathbf{L}^q(x, \xi) \quad (2.31)$$

where

$$\Delta \mathbf{K}^q(x, \xi), \Delta \mathbf{L}^q(x, \xi) \leq C \frac{M^q (x - (1 - \varepsilon)\xi)^q}{q!}. \quad (2.32)$$

in which $C, M > 0$.

Improved successive approximation method (LH, F. DiMeglio, R. Vazquez and M. Krstic (2015 a))



Key point:

The proof is based on the fact that, starting from any point (x, ξ) , all the characteristic lines "get closer" to the line defined by $x - (1 - \epsilon)\xi = 0$.

Since $\beta(t, 1) = 0$ and

$$\begin{pmatrix} \alpha(t, x) \\ \beta(t, x) \end{pmatrix} = \begin{pmatrix} u(t, x) \\ v(t, x) \end{pmatrix} - \int_0^x \begin{pmatrix} 0 & 0 \\ K(x, \xi) & L(x, \xi) \end{pmatrix} \begin{pmatrix} u(t, \xi) \\ v(t, \xi) \end{pmatrix} d\xi. \quad (2.33)$$

Our feedback laws finally is

$$U(t) = \int_0^1 [K(1, \xi)u(\xi) + L(1, \xi)v(\xi)] d\xi - R_1 u(t, 1). \quad (2.34)$$

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Theorem: Finite-time stabilization (LH, F. DiMeglio, R. Vazquez and M. Krstic (2015 a))

By $U(t)$, the zero equilibrium of the original system is reached in finite time $t = t_F$.

Proof. Good! (2.34) is always invertible, i.e. there exists a matrix function $\mathcal{R} \in (L^\infty(\mathcal{T}))^{(n+m) \times (n+m)}$, such that

$$\begin{pmatrix} u(t, x) \\ v(t, x) \end{pmatrix} = \begin{pmatrix} \alpha(t, x) \\ \beta(t, x) \end{pmatrix} - \int_0^x \mathcal{R}(x, \xi) \begin{pmatrix} \alpha(t, \xi) \\ \beta(t, \xi) \end{pmatrix} d\xi. \quad (2.35)$$

Since $(\alpha, \beta)^T$ goes to zero in finite time $t = t_F$, therefore $(u, v)^T$ shares this property.

- [1] We can deal with the observer problem by using also the backstepping approach.
- [2] Our method is still valid for the case if the coefficients $(\Lambda^\pm, \Sigma^{\pm, \pm})$ are depending on x .
- [3] Counter Example: G.Bastin-JMC-2016.

$$S_1(t, 0) = \int_0^L K(0, \xi) S_1(t, \xi) + L(0, \xi) S_2(t, \xi) d\xi \quad (2.36)$$

- [5] Our kernel is probably not continuous, which is different with previous works.

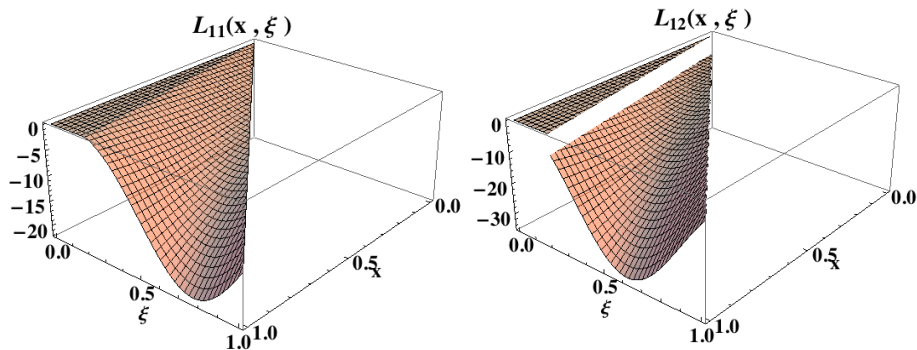


Figure 3 : Kernels $L_{11}(x, \xi)$ and $L_{12}(x, \xi)$ ($n = 0, m = 2$).

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- 4 Perspectives-Stabilization of nonlocal hyperbolic system

The system considered is

$$\frac{\partial u}{\partial t} + A(x, u) \frac{\partial u}{\partial x} = F(x, u) \quad (t, x) \in [0, T] \times [0, L], \quad (3.1)$$

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$$\frac{\partial u}{\partial t} + A(x, u) \frac{\partial u}{\partial x} = F(x, u) \quad (t, x) \in [0, T] \times [0, L], \quad (3.1)$$

where

- $u = (u_1, \dots, u_n)^T$ is a vector function of (t, x) ;
- $A(x, u) := (a_{ij}(x, u))_{n \times n}$ is of class C^2 , $A(x, 0)$ is a diagonal matrix with distinct and nonzero eigenvalues $A(x, 0) = \text{diag}(\Lambda_1(x), \dots, \Lambda_n(x))$, which are ordered as follows:

$$\Lambda_1(x) < \Lambda_2(x) < \dots < \Lambda_m(x) < 0 < \Lambda_{m+1}(x) < \dots < \Lambda_n(x), \forall x \in [0, 1]. \quad (3.2)$$

System Description

- $F : [0, 1] \times \mathbb{R}^n \rightarrow \mathbb{R}^n$ is a vector valued function with C^2 components $f_i(x, u)$ ($i = 1, \dots, n$) with respect to u and

$$F(x, 0) \equiv 0. \quad (3.3)$$

- Denote

$$\frac{\partial F}{\partial u}(x, 0) := (f_{ij}(x))_{n \times n}, \quad (3.4)$$

we assume that $f_{ij} \in C^2([0, 1])$ and

$$f_{ii}(x) \equiv 0. \quad (3.5)$$

Remark

One can make some coordinate transformations in order to guarantee that (3.5) is valid (see JMC (2014) and L. Hu & F. DiMeglio (2015)).

The boundary conditions are given as follows:

$$x = 0 : u_s = G_s(u_1, \dots, u_m), \quad s = m + 1, \dots, n, \quad (3.6)$$

$$x = 1 : u_r = h_r(t), \quad r = 1, \dots, m, \quad (3.7)$$

where G_s are C^2 functions, and we assume that they vanish at the origin, i.e.

$$G_s(0, \dots, 0) \equiv 0, \quad s = m + 1, \dots, n, \quad (3.8)$$

while $H = (h_1, \dots, h_m)^T$ are boundary controls.

Remark

- Local well-posedness: see JMC (2008) and Tatsien Li (1994) [Remark 1.3 on page 171] etc.

Theorem (LH, R. Vazquez, F. Di Meglio and M. Krstic (2015 b))

For every $\lambda > 0$, there exist $\delta > 0$, $c > 0$ and a continuous linear feedback control H , such that 0 is the exponential stable point of $u = u(t, x)$, i.e.

$$\|u(t, \cdot)\|_{H^2} \leq ce^{-\lambda t} \|u(0, \cdot)\|_{H^2}, \quad (3.9)$$

provided that $\|u(0, \cdot)\|_{H^2} \leq \delta$.

Remark

- For simplicity, we skip of C^1 compatibility conditions at the points $(t, x) = (0, 0)$ and $(0, 1)$;

[1] Linearized system is

$$\begin{cases} u_t + \Lambda(x)u_x = \Sigma(x)u \\ x = 0 : u_s(t, 0) = \sum_{j=1}^m Q_{sj}u_j(t, 0) \\ x = L : u_j(t, 1) = h_j(t). \end{cases} \quad (3.10)$$

where $\Lambda(x) = A(x, 0)$ and $\Sigma(x) = \frac{\partial F}{\partial u}(x, 0)$, then we can find

$$h_j(t) = \int_0^1 \sum_{l=1}^n K_{jl}(1, \xi)u_l(t, \xi)d\xi, \quad (j = 1, \dots, m), \quad (3.11)$$

which can stabilize the linearized system in finite-time.

Sketch of the Proof

[2] We also use the volterra transformation

$$\gamma(t, x) = u(t, x) - \int_0^x K(x, \xi)u(t, \xi)dt. \quad (3.12)$$

Lemma:Regularity of the direct kernel (LH, R. Vazquez, F. DiMeglio and M. Krstic (2015 b))

Let $N \in \mathbb{N}^+$. Under the assumption that $\sigma_{ij} \in C^N[0, 1]$, $\lambda_i \in C^N[0, 1](i, j = 1, \dots, n)$, there exists a unique piecewise $C^N(\mathcal{T})$ solution to K kernel. Moreover, then $K(\cdot, \cdot) \in C^{N-1}(0, 1)$, $K(\cdot, 0) \in C^{N-1}(0, 1)$ with bounded C^{N-1} norm.

Remark

- The H^2 norm of γ is equivalent to the H^2 norm of u .

Then the nonlinear target system is

$$\begin{aligned} & \gamma_t(t, x) + \Lambda(x)\gamma_x(t, x) - G(x)\gamma(t, 0) \\ & = F_3[\gamma, \gamma_x] + F_4[\gamma], \end{aligned} \quad (3.13)$$

The boundary conditions are

$$x = 0 : \gamma_+(t, 0) = Q\gamma_-(t, 0) + G_{NL}(\gamma_-(t, 0)) \quad (3.14)$$

and

$$x = 1 : \gamma_-(t, 1) = 0. \quad (3.15)$$

- [3] Luckily! The usual Lyapunov function (see JMC-R.Vazquez-M.Krstic -G.Bastin, SICON 2013) can be also used to exponentially stabilize this γ system.

Control Lyapunov Functions

- Estimate of $\|\gamma\|_{L^2}$

Define

$$V_1(t) = \int_0^1 e^{-\delta x} \gamma_+(t, x)^T (\Lambda_+(x))^{-1} \gamma_+(t, x) dx \\ - \int_0^1 e^{\delta x} \gamma_-(t, x)^T B (\Lambda_-(x))^{-1} \gamma_-(t, x) dx.$$

We have

Proposition 1

For any given $\lambda_1 > 0$, there exists $\delta_1 > 0$ and $K_2 > 0$, such that

$$\dot{V}_1 \leq -\lambda_1 V_1 + K_2 (V_1^{\frac{3}{2}} + \|\gamma_x\|_{\infty} V_1), \quad (3.16)$$

provided $\|\gamma\|_{\infty} \leq \delta_1$.

Control Lyapunov Functions

- Estimate of $\|\gamma_t\|_{L^2}$

Define $\zeta = \gamma_t$ and

$$V_2(t) = \int_0^1 \zeta^T(t, x) R[\gamma] \zeta(t, x) dx, \quad (3.17)$$

where $R[\gamma]$ is a positive matrix. We have

Proposition 2

For any given $\lambda_2 > 0$, there exists $\delta_2 > 0$ and $K_7 > 0$, such that

$$\dot{V}_2 \leq -\lambda_2 V_2 + K_7 (\|\zeta\|_\infty + \|\gamma\|_\infty) V_2, \quad (3.18)$$

provided that $\|\gamma\|_\infty \leq \delta_2$.

Control Lyapunov Functions

- Estimate of $\|\gamma_{tt}\|_{L^2}$
Define $\theta = \gamma_{tt}$ and

$$V_3(t) = \int_0^1 \theta^T(t, x) R[\gamma] \theta(t, x) dx, \quad (3.19)$$

We have

Proposition 3

For any given $\lambda_3 > 0$, there exists $\delta_3 > 0$ and positive constants K_{10} , K_{11} , K_{12} , K_{13} and K_{14} , such that

$$\dot{V}_3 \leq -\lambda_3 V_3 + K_{10} \|\gamma\|_\infty V_3 + K_{11} V_3 V_2^{\frac{1}{2}} + K_{12} V_2 V_3^{\frac{1}{2}} + K_{13} V_3^{\frac{3}{2}} + K_{14} \|\zeta\|_\infty^3, \quad (3.20)$$

provided that $\|\gamma\|_\infty + \|\zeta\|_\infty \leq \delta_3$.

Proof of main result

Denote $W = V_1 + V_2 + V_3$, by Proposition 1–3, one can show that for any given $\lambda > 0$, there exists $\delta > 0$ and $K_{15} > 0$, such that

$$\dot{W} \leq -\lambda W + K_{15} W^{\frac{3}{2}}, \quad (3.21)$$

provided that $\|\gamma\|_\infty + \|\zeta\|_\infty \leq \delta$.

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Non-local hyperbolic system is considered as

$$\begin{cases} u_t = u_x + \int_0^L g(x, y)u(t, y)dy \\ u(t, L) = U(t) \end{cases} \quad (4.1)$$

- u is a scalar $U(t)$ is the boundary feedback.
- (4.1) may involve the traffic laws, some KdV-like equation and PDE-ODE interconnected system (see M. Krstic, A.Smyshlyaev (2008), F. Bribiesca and M. Krstic (2015)).

- M. Krstic, A.Smyshlyaev (2008): $g(x, y) = 0, x \leq y$.
Map (4.1) into

$$\begin{cases} w_t = w_x \\ w(t, L) = 0. \end{cases} \quad (4.2)$$

by using Volterra transformation

$$u(t, x) = w(t, x) - \int_0^x k(x, y)w(t, y)dy \quad (4.3)$$

- F. Bribiesca and M. Krstic (2015): For general g Volterra transformation fails. Map (4.1) into (4.2) by using Fredholm transformation

$$u(t, x) = w(t, x) - \int_0^L \tilde{k}(x, y)w(t, y)dy \quad (4.4)$$

provided $\|g\|$ is small.

Remark

The smallness of $\|g\|$ is used to guarantee the existence of \tilde{k} and the invertibility of the the Fredholm transformation (4.4).

Main results (J.-M. Coron, LH and G. Olive (2015))

Suppose $g \in H^1(\mathcal{T}_-) \cap H^1(\mathcal{T}_+)$, where

$$\mathcal{T}_- = \{(x, y) \in (0, L) \times (0, L) | x > y\},$$

$$\mathcal{T}_+ = \{(x, y) \in (0, L) \times (0, L) | x < y\},$$

Then (4.1) is finite-time stabilizable in time L if and only if (4.1) is exactly controllable at time L .

We also use the Fredholm transformation

$$u(t, x) = w(t, x) - \int_0^L h(x, y)w(t, y)dy \quad (4.5)$$

to map the system (4.2) into (4.1). The kernel h satisfies

$$\begin{cases} h_x(x, y) + h_y(x, y) + \int_0^L G(x, \tau)h(\tau, y)d\tau - G(x, y) = 0 \\ h(x, 0) = h(x, L) = 0 \end{cases} \quad (4.6)$$

- We use the controllability of the original system (4.1) to prove the existence of h ;
- The invertibility of h can be guaranteed if the original system (4.1) is controllability.

- We can also treat the equation of the more general form

$$\begin{cases} u_t(t, x) = u_x(t, x) + \alpha(x)u(t, x) + \beta(x)u(t, 0) + \int_0^L g(x, y)u(t, y)dy \\ u(t, L) = \int_0^L \gamma(x)u(t, x)dx + U(t) \\ u(0, x) = u^0(x). \end{cases}$$

where $\alpha, \beta, \gamma : (0, L) \rightarrow \mathbb{C}$ are regular enough.

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$$\begin{cases} u_t(t, x) = u_x(t, x) + \alpha(x)u(t, x) + \beta(x)u(t, 0) + \int_0^L g(x, y)u(t, y)dy \\ u(t, L) = \int_0^L \gamma(x)u(t, x)dx + U(t) \\ u(0, x) = u^0(x). \end{cases}$$

where $\alpha, \beta, \gamma : (0, L) \rightarrow \mathbb{C}$ are regular enough.

- We expect that (4.1) is always controllable. However, this is NOT true.
 - If $g(x, y) = 0, x \leq y$, (4.1) is controllable. (M. Krstic, A.Smyshlyaev (2008))
 - If $\|g\|$ is small enough, (4.1) is controllable. (F. Bribiesca and M. Krstic (2015))
 - If $g(x, y) = g(x)$, (4.1) is controllable if and only if (J.M.Coron, LH and G.Olive (2015))

$$\ker(\lambda - A^*) \cap \ker B^* = \{0\}, \quad \forall \lambda \in \mathbb{C}. \quad (4.7)$$

Question: What about the general case $g(x, y)$?

Pas un jour sans contrôle

Thank you for your attention !