

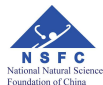
A method to determine the minimal null control time of 1D linear hyperbolic systems

Guillaume Olive

(joint work with Long Hu)

Control of PDEs and related topics

Toulouse, July 2, 2025



Outline of the talk

I. Framework

II. Preliminary results

III. Presentation of the method

System description

Equations

$$\frac{\partial y}{\partial t}(t, x) + \Lambda(x) \frac{\partial y}{\partial x}(t, x) = M(x)y(t, x), \quad (t, x) \in R_T,$$

- $R_T = (0, T) \times (0, 1)$ and $y : R_T \rightarrow \mathbb{R}^n$ is the state.
- $\Lambda \in C^{0,1}([0, 1])^{n \times n}$ is diagonal, with

$$\lambda_1 < \dots < \lambda_m < 0 < \lambda_{m+1} < \dots < \lambda_{m+p}.$$

- $M \in L^\infty(0, 1)^{n \times n}$ is the **internal coupling matrix**.

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$$y(0, x) = y^0(x).$$

Denoting by $y = \begin{pmatrix} y_- \\ y_+ \end{pmatrix} \in \mathbb{R}^{m+p}$,

Boundary conditions

$$y_-(t, 1) = u(t), \quad y_+(t, 0) = Qy_-(t, 0).$$

$u : (0, T) \rightarrow \mathbb{R}^m$ is the **control**. $Q \in \mathbb{R}^{p \times m}$ is the **boundary coupling matrix**.

Minimal control time

Well-posedness: $\forall y^0 \in L^2, \forall u \in L^2, \exists! y \in C^0([0, T]; L^2(0, 1)^n)$.

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Remark: **(NC)** in time $T_1 \implies$ **(NC)** in time $T_2 \geq T_1$.

Definition

Minimal time for **(NC)**:

$$T_{\text{inf}} = \inf \{ T > 0, \text{ System is (NC) in time } T \}, \quad (\in [0, +\infty]).$$

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Goal

$$T_{\text{inf}} = ??? \quad (Q, M \text{ are fixed}).$$

Control of a single equation

The transport equation:

$$\begin{cases} \frac{\partial y}{\partial t} + \lambda(x) \frac{\partial y}{\partial x} = 0, \\ y(0, x) = y^0(x), \\ y(t, \delta) = u(t), \end{cases} \quad (\text{TE})$$

($\delta = 1$ if $\lambda < 0$, $\delta = 0$ if $\lambda > 0$). We have

$$T_{\text{inf}} = \int_0^1 \frac{1}{|\lambda(\xi)|} d\xi.$$

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Coming back to the system,

Definition

$T_k = T_{\text{inf}}$ of (TE) with speed λ_k .

$$\lambda_1 < \dots < \lambda_m < 0 < \lambda_{m+1} < \dots < \lambda_{m+p},$$

$$\text{implies } T_1 \leq \dots \leq T_m, \quad T_{m+1} \geq \dots \geq T_{m+p}.$$

Some known results

Assumptions	Results	References
\emptyset	(NC) in time $T_{m+1} + T_m$	Russell 1978; Li and Rao 2003
$M = 0$	Explicit formula for T_{inf}	Weck 1982; Hu and Olive 2022
$Q \in \mathcal{B}$	(NC) in any time $T > T_{[\text{CN}]}$	Coron and Nguyen 2019, 2021
$\text{rank } Q = p$	T_{inf} is the same as for $M = 0$	Hu and Olive 2021a, 2022
$m = p = 1, Q \neq 0$	$T_{\text{inf}} = T_2 + T_1$	Coron, Vazquez, et al. 2013
$Q = 0$	Explicit formula for T_{inf}	Hu and Olive 2021b
\emptyset	$\min_M T_{\text{inf}}$ & $\max_M T_{\text{inf}}$	Hu and Olive 2022
First $\min \{m, p\}$ rows of Q lin. indep.	T_{inf} is the same as for $M = 0$	
$m = 1, M(x) = M, \text{diag } M = 0$	Explicit formula for T_{inf}	Hu and Olive 2025b

Still, T_{inf} remains unknown in some situations.

We propose a method when Λ, M are smooth enough.

Preliminary step 1: Backstepping method for PDEs

Assume $\text{diag } M = 0$ (for simplicity). Volterra transformation (Krstic and Smyshlyaev 2008):

$$\tilde{y}(t, x) = y(t, x) - \int_0^x K(x, \xi) y(t, \xi) d\xi,$$

where $K : \{0 < \xi < x < 1\} \rightarrow \mathbb{R}^{n \times n}$ is a kernel (to be determined).

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Equivalent system

$$\begin{cases} \frac{\partial \tilde{y}}{\partial t} + \Lambda(x) \frac{\partial \tilde{y}}{\partial x} = \mathbf{G}(x) \tilde{y}_-(t, 0), \\ \tilde{y}(0, x) = \dots, \\ \tilde{y}_-(t, 1) = \dots, \quad \tilde{y}_+(t, 0) = \mathbf{Q} \tilde{y}_-(t, 0), \end{cases}$$

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$$\begin{cases} \frac{\partial \tilde{y}}{\partial t} + \Lambda(x) \frac{\partial \tilde{y}}{\partial x} = G(x) \tilde{y}_-(t, 0), \\ \tilde{y}(0, x) = \dots, \\ \tilde{y}_-(t, 1) = \dots, \quad \tilde{y}_+(t, 0) = Q \tilde{y}_-(t, 0), \end{cases}$$

■ where $G(x) = K(x, 0)B$, $B = -\Lambda(0) \begin{pmatrix} \text{Id}_m \\ Q \end{pmatrix}$,

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■ where $G(x) = K(x, 0)B$, $B = -\Lambda(0) \begin{pmatrix} \text{Id}_m \\ Q \end{pmatrix}$,

■ provided that K satisfies the **kernel equations**:

$$\begin{cases} \Lambda(x) \frac{\partial K}{\partial x}(x, \xi) + \frac{\partial K}{\partial \xi}(x, \xi) \Lambda(\xi) + K(x, \xi) (\Lambda'(\xi) + M(\xi)) = 0, \\ \Lambda(x) K(x, x) - K(x, x) \Lambda(x) = M(x). \end{cases}$$

Such K exist (Hu, Di Meglio, et al. 2016).

Preliminary step 2: Compactness-uniqueness method

Denote by $G = \begin{pmatrix} G_{--} \\ G_{+-} \end{pmatrix}$, where $G_{--} =$ the first m rows of G .

Theorem (Hu and Olive 2022)

T_{inf} does not depend on G_{--} .

Systems with $G_{--} = 0$ will be denoted by

$$(Q, G_{+-}).$$

Remarks:

- Impossible to obtain by transformation (e.g. Backstepping).
- We are dealing with **(NC)**, not **(EC)**.

Ingredient 1: T_{inf} for a class of full rank Q

Let $n^* = \min \{m, p\}$.

Theorem (Hu and Olive 2022)

Assume that the *first n^* rows of Q are lin. indep.* Then,

$$T_{\text{inf}}(Q, G_{+-}) = T_{\text{inf}}(Q, \mathbf{0}).$$

Ingredient 2: Linear combination of rows

Below, $Q_j = j$ -th row of Q , and $G_j = j$ -th row of G_{+-} .

Lemma (Hu and Olive 2022)

Let $a_1, \dots, a_{k-1} \in \mathbb{R}$. We have

$$T_{\text{inf}}(Q, G_{+-}) = T_{\text{inf}}(\tilde{Q}, \tilde{G}_{+-}),$$

where:

$$\tilde{Q} = \begin{pmatrix} Q_1 \\ \vdots \\ Q_{k-1} \\ \hline Q_k - \sum_{j < k} a_j Q_j \\ \hline Q_{k+1} \\ \vdots \\ Q_p \end{pmatrix}, \quad \tilde{G}_{+-}(x) = \begin{pmatrix} G_1(x) \\ \vdots \\ G_{k-1}(x) \\ \hline G_k(x) - \sum_{j < k} a_j G_j(\zeta_{kj}(x)) \\ \hline G_{k+1}(x) \\ \vdots \\ G_p(x) \end{pmatrix},$$

where ζ_{kj} is the solution to the ODE: $\zeta'_{kj} = \frac{\lambda_{m+j}(\zeta_{kj})}{\lambda_{m+k}}$ with $\zeta_{kj}(0) = 0$.

Ingredient 3: Derivative of zero rows

Lemma (Hu and Olive 2025a)

Assume that $Q_k = 0$. Then,

$$T_{\text{inf}}(Q, G_{+-}) = T_{\text{inf}}(\hat{Q}, \hat{G}_{+-}),$$

where:

$$\hat{Q} = \begin{pmatrix} Q_1 \\ \vdots \\ Q_{k-1} \\ \mathbf{G}_k(0) \\ Q_{k+1} \\ \vdots \\ Q_p \end{pmatrix}, \quad \hat{G}_{+-}(x) = \begin{pmatrix} G_1(x) \\ \vdots \\ G_{k-1}(x) \\ (\mathcal{D}_k G_k)(x) \\ G_{k+1}(x) \\ \vdots \\ G_p(x) \end{pmatrix},$$

where $\mathcal{D}_k = \lambda_{m+k} \frac{d}{dx}$.

Presentation of the method

Main ideas

Construct systems $(Q^{[k]}, G_{+-}^{[k]})$, $k = 1, 2, \dots, n^*$, such that:

- $T_{\text{inf}}(Q^{[k]}, G_{+-}^{[k]}) = T_{\text{inf}}(Q^{[k-1]}, G_{+-}^{[k-1]})$.
- First k rows of $Q^{[k]}$ are lin. indep.

If we succeed,

$$T_{\text{inf}}(Q, G_{+-}) = T_{\text{inf}}(Q^{[n^*]}, G_{+-}^{[n^*]}) = T_{\text{inf}}(Q^{[n^*]}, \mathbf{0}).$$

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Construction:

- First k rows of $Q^{[k-1]}$ are lin. indep.: OK.
- Otherwise,
 - ▶ **Step 1: Put the k -th row to zero** (ingredient 2):

$$\text{new } Q_k: Q_k^{[k-1]} - \sum_{j < k} a_j Q_j^{[k-1]} = 0, \quad \text{new } G_k: G_k^{[k-1]}(x) - \sum_{j < k} a_j G_j^{[k-1]}(\zeta_{kj}(x)).$$

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- ▶ **Step 2: Take the derivative of the k -th row** (ingredient 3):

$$\text{new } Q_k: G_k^{[k-1]}(0) - \sum_{j < k} a_j G_j^{[k-1]}(0), \quad \text{new } G_k: (\mathcal{D}_k G_k^{[k-1]})(x) - \sum_{j < k} a_j (\mathcal{D}_k (G_j^{[k-1]} \circ \zeta_{kj}))(x).$$

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Then repeat! **Main assumption:** We eventually find a lin. indep. new row.

Computations

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Solution: recall $G(x) = K(x, 0)B$ and use the kernel equations

$$\begin{cases} \Lambda(x) \frac{\partial K}{\partial x}(x, \xi) + \frac{\partial K}{\partial \xi}(x, \xi) \Lambda(\xi) + K(x, \xi) (\Lambda'(\xi) + M(\xi)) = 0, \\ \Lambda(x) K(x, x) - K(x, x) \Lambda(x) = M(x). \end{cases}$$

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Indeed ($\Lambda = \text{cst}$ for simplicity),

■ 0-th order:

$$k_{ij}(0, 0) = \frac{m_{ij}(0)}{\lambda_i - \lambda_j}, \quad (i \neq j).$$

■ 1-st order:

$$\begin{pmatrix} \lambda_i & \lambda_j \\ 1 & 1 \end{pmatrix} \begin{pmatrix} \frac{\partial k_{ij}}{\partial x}(0, 0) \\ \frac{\partial k_{ij}}{\partial \xi}(0, 0) \end{pmatrix} = \begin{pmatrix} -\sum_{r=1}^n k_{ir}(0, 0) m_{rj}(0) \\ \frac{m'_{ij}(0)}{\lambda_i - \lambda_j} \end{pmatrix}, \quad (i \neq j).$$

■ Etc.

A 3 + 2 constant system

$$\Lambda = \text{diag} \left(-2, -1, -\frac{1}{2}, 1, 2 \right) \quad M = \left(\begin{array}{ccc|cc} 0 & 0 & 0 & 1 & 6 \\ 0 & 0 & 0 & 2 & 1 \\ 0 & 0 & 0 & 3 & -1 \\ \hline 3 & 2 & 1 & 0 & 0 \\ 8 & 9 & -\frac{20}{3} & 0 & 0 \end{array} \right), \quad Q = \begin{pmatrix} 0 & 1 & -2 \\ 0 & 2 & -4 \end{pmatrix}.$$

We have

$$T_1 = \frac{1}{2}, \quad T_2 = 1, \quad T_3 = 2, \quad T_4 = 1, \quad T_5 = \frac{1}{2}.$$

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Known results:

- Russell 1978: **(NC)** in time $T = T_4 + T_3 = 3$.
- Coron and Nguyen 2019, 2021: not applicable ($Q \notin \mathcal{B}$).
- Hu and Olive 2022:

$$\min_M T_{\text{inf}} = \max \{ T_4 + T_2, \quad T_3 \} = 2, \quad \max_M T_{\text{inf}} = \max \{ T_4 + T_2, \quad T_5 + T_3 \} = \frac{5}{2}.$$

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With our method:

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With our method:

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Thank you for your attention !

More details at:

<https://doi.org/10.1016/j.jde.2025.113455>