Computer-assisted proofs of non-reachability for linear parabolic control systems

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Outline



2 Method & results

- Separation argument
- Discretisation & error control
- 3 Numerical applications



Outline

Context & objectives

2 Method & results

- Separation argument
- Discretisation & error control

3 Numerical applications

4 Conclusion

Context	&	objectives
0000		

Numerical applications

Conclusion

	1.2					
$\forall t, x \in [0, T] \times [0, 1],$	0.8 -					
$\int \dot{y}(t,x) = \Delta y(t,x)$	0.6 -					
$\begin{cases} y(0, x) = 0\\ y(t, 0) = y(t, 1) = 0. \end{cases}$	0.4 -					
	0.2 -					
	0.0 -					
	-0.2	0.2	0.4	0.6	0.8	1.0

Context	&	objectives
0000		

Numerical applications

Conclusion



Context	&	objectives
0000		

Numerical applications

Conclusion

$$\forall t, x \in [0, T] \times [0, 1],$$

$$\begin{cases} \dot{y}(t, x) = \Delta y(t, x) + u(t, x) \\ y(0, x) = y_0(x) = 0 \\ y(t, 0) = y(t, 1) = 0 \\ u(t, x) \in \mathbb{R}. \end{cases}$$

Context	&	ob	ectives
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Numerical applications

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Motivation

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Context	&	ob	ectives
0000			

Numerical applications

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Context	&	ob	ect	ives
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Numerical applications



Context	&	ob	ectives
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Numerical applications



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Control system

We call linear control system

$$\begin{cases} \dot{y}(t) = Ay(t) + Bu(t) \quad \forall t \in [0, T] \\ y(0) = y_0 \\ u(t) \in \mathcal{U}_0 \qquad \forall t \in [0, T]. \end{cases}$$

Which allows the Duhamel decomposition

$$y(T,\cdot;y_0,u)=S_Ty_0+L_Tu.$$

We call the constraint set

$$\mathcal{U} = \left\{ \boldsymbol{u}, \quad \forall t \in [0, T], \ \boldsymbol{u}(t) \in \mathcal{U}_0 \right\}.$$

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Reachability

Definition

A target y_f is \mathcal{U} -reachable in time T if :

$$\exists u \in \mathcal{U}, y(T, \cdot; u) = y_f.$$

We call reachable set and denote $L_T \mathcal{U}$ the set of all \mathcal{U} -reachable points.



Outline

Context & objectives

2

Method & results

- Separation argument
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3 Numerical applications

4 Conclusion

Numerical applications

Conclusion

Non-reachability : Geometric intuition





 $L_T \mathcal{U}$

Numerical applications

Conclusion

Non-reachability : Geometric intuition



 p_f



Numerical applications

Conclusion

Non-reachability : Geometric intuition



Numerical applications

Conclusion

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Numerical applications

Conclusion

Non-reachability : theorem

Denoting

$$J: p_f \mapsto \sigma_{L_T \mathcal{U}}(p_f) - \langle p_f, y_f \rangle,$$

where

$$\sigma_{L_{\mathcal{T}}} : p_f \mapsto \sup_{x \in L_{\mathcal{T}}} \langle x, p_f \rangle.$$

Theorem

Numerical applications

Conclusion

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Numerical applications

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Numerical applications

Conclusion

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Theorem

Context & objectives	Method & results	Numerical applications	Conclusion
Method			

Theorem

Context & objectives	Method & results ○○○●○○○○	Numerical applications	Conclusion
Method			

Theorem

If there exists p_f such that $J(p_f) < 0$, then y_f is not \mathcal{U} -reachable for (*S*) in time *T*.

To use this theorem to prove the non-reachability of y_f , three steps are required :

- discretise J into $J_{\Delta t,h} \simeq J$ such that we can evaluate $J_{\Delta t,h}$
- 3 find p_{fh} such that $J_{\Delta t,h}(p_{fh}) < 0$
- 3 associate p_{fh} to some p_f and check that $J(p_f) < 0$

Context & objectives	Method & results	Numerical applications	Conclusion
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 - interpolate p_{fh} into p_f
 - bound discretisation errors e_d(p_f)
 - bound round-off errors $e_r(p_f)$.
 - check that $J_{\Delta t,h}(p_f) + e_d(p_f) + e_r(p_f) < 0.$

Context	objectives
0000	

Numerical applications

Conclusion

Computing *J*

J can be reformulated as :

$$J: \mathbf{p}_{f} \mapsto \int_{0}^{T} \sigma_{\mathcal{U}_{0}}(\mathbf{B}^{*} S_{T-t}^{*} \mathbf{p}_{f}) \, \mathrm{d}t - \langle \mathbf{p}_{f}, \mathbf{y}_{f} \rangle,$$

where we assume $\sigma_{\mathcal{U}_0}$ has a known closed-form expression.

Context	objectives
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Numerical applications

Conclusion

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Time discretisation – Implicit Euler scheme :

$$J_{\Delta t}(\boldsymbol{p}_{f}) = \Delta t \sum_{n=0}^{N_{t}-1} \sigma_{\mathcal{U}_{0}}(\boldsymbol{B}^{*}(\mathsf{Id} - \Delta t\boldsymbol{A}^{*})^{-(Nt-n)}\boldsymbol{p}_{f}) - \langle \boldsymbol{p}_{f}, \boldsymbol{y}_{f} \rangle,$$

Context	objectives
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Numerical applications

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Space discretisation - P1 finite elements :

$$J_{\Delta t,h}(\boldsymbol{p}_{fh}) = \Delta t \sum_{n=0}^{N_t-1} \sigma_{\mathcal{U}_0}(B_h^*(\mathrm{Id} - \Delta t A_h^*)^{-(Nt-n)} \boldsymbol{p}_{fh}) - \langle \boldsymbol{p}_{fh}, \boldsymbol{y}_{fh} \rangle.$$

Numerical applications

Hypotheses on A

Let A verify :

Numerical applications

Conclusion

Hypotheses on A

Let A verify :

• M α -accretivity : for $0 < \alpha < \frac{\pi}{2}$,

 $\forall v \in \mathcal{D}(A), \quad \langle Av, v \rangle \in \mathcal{S}_{\alpha} \text{ and,}$

 $\forall z \notin \mathcal{S}_{\alpha}, z \mid -A \text{ is an isomorphism}$ from $\mathcal{D}(A)$ to H.



Numerical applications

Conclusion

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• Coercivity : for $a_1 > 0$

 $\forall \mathbf{v} \in \mathcal{D}(\mathbf{A}), \quad \mathsf{Re}\langle \mathbf{A}\mathbf{v}, \mathbf{v} \rangle \geq a_1 \|\mathbf{v}\|^2.$



Numerical applications

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• Continuity : for $0 < a_0 \le a_1$,

 $\forall \mathbf{v}, \mathbf{w} \in \mathcal{D}(\mathbf{A}) \times \mathbf{V}, \quad |\langle \mathbf{A}\mathbf{v}, \mathbf{w} \rangle| \leq a_0 \|\mathbf{v}\| \|\mathbf{w}\|.$





Discretisation error

Theorem

Let $p_{fh} \in V_h$ and $p_f \in \mathcal{D}(A^*)$ such that $\forall i \in \{0, ..., N_x\}$, $p_{fh}(ih) = p_f(ih)$. Then, if A satisfy the previously stated hypotheses, we have

 $e_d(p_f) = |J(p_f) - J_{\Delta t,h}(p_{fh})| \le (C_1 \Delta t + C_2 h^2) \|A^* p_f\|$

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$$\boldsymbol{e}_{d}(\boldsymbol{p}_{f}) = |J(\boldsymbol{p}_{f}) - J_{\Delta t,h}(\boldsymbol{p}_{fh})| \leq (C_{1}\Delta t + C_{2}h^{2}) \|\boldsymbol{A}^{*}\boldsymbol{p}_{f}\|,$$

with

$$\begin{aligned} C_1 &= \frac{1}{2} MT \|B\| + \frac{C_{\alpha}}{\cos(\alpha)} \Big(\|y_0\|_X + MT \|B\| \Big) \\ C_2 &= \left(\frac{a_1^2 C_0^2}{a_0} \left(7 + \frac{4\ln(2)}{\cos(\alpha)} + C_{\alpha} \right) + C_{\alpha} C_0 \right) \Big(\|y_0\|_X + MT \|B\| \Big) \\ C_{\alpha} &\leq 2 + \frac{2}{\sqrt{3}} \qquad \& \qquad C_0 = \frac{1}{2}. \end{aligned}$$



The Intlab library, encoded in Matlab by Siegfried M. Rump, takes care of it for us.

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Outline

Context & objectives

2 Method & results

- Separation argument
- Discretisation & error control

3 Numerical applications

4 Conclusion

Context	objectives
0000	

Numerical applications

Numerical example

Back to the original example : $\forall t, x \in [0, T] \times [0, 1]$

$$\begin{cases} \dot{y}(t,x) = \Delta y(t,x) + \mathbb{1}_{\omega} u(t,x) \\ y(0,x) = y_0(x) = 0 \\ y(t,0) = y(t,1) = 0 \\ 0 \le u(t,x) \le 1 \\ y(T,x) = y_f(x) = \sin(\pi x). \end{cases}$$



Context	objectives
0000	

Numerical applications

Conclusion

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$$\begin{cases} \dot{y}(t,x) = 0.1 \Delta y(t,x) + \mathbb{1}_{\omega} u(t,x) \\ y(0,x) = 0 \\ y(t,0) = y(t,1) = 0 \\ 0 \le u(t,x) \le 1 \\ y(T,x) = y_f(x) = 0.21 \sin(\pi x). \end{cases}$$



Context	objectives
0000	

Numerical applications

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Back to the original example :

 $y_f = 0.21 \sin(\pi \cdot).$

Is $y_f \mathcal{U}$ -reachable in time T = 1?



Context	objectives
0000	

Numerical applications

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Numerical applications

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Proposition

 y_f is not \mathcal{U} -reachable in time T = 1. Indeed,

$$J(p_f) \in [-0.0093, -0.0035] < 0.$$



Numerical applications

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What is a guaranteed lower-bound of the minimal time of reachability t^* of y_f ?



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Cubic spline i	nterpolation		

 $\forall i \in \{0, \ldots, N_x\}, \quad p_f(ih) = p_{fh}(ih).$

For a given p_{fh} P1 element, we need to find $p_f \in \mathcal{D}(A)$ such that

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Context & objectives	Method & results	Numerical applications	Conclusion

Cubic spline interpolation

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Lemma

If $A = \Delta$, cubic splines are the optimal way to interpolate p_{fh} , in the sense that

$$\inf_{\substack{p_f \in \mathcal{D}(\Delta)}} \|\Delta p_f\| = \inf_{\substack{p_f \in \mathcal{D}(\Delta)\\p_f \text{ cubic spline}}} \|\Delta p_f\|.$$

Furthermore, the inf is reached, the optimal spline has a closed-form expression satisfying $p_f \in C^2([0,1])$.

Context & objectives	Method & results	Numerical applications	Conclusion

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Outline

Context & objectives

2 Method & results

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3 Numerical applications



Conclusion and perspectives

Contributions :

- A general method to analyse the non-reachability of targets of linear control problems
- Fine explicit estimates for a wide class of parabolic control problems

Perspectives :

- Apply the method for other classes of linear PDEs
- For ODEs, develop a method to prove numerically the reachability of a given target and approximate the reachable set with guaranteed sets (*work in progress*)