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Recovering the initial state of a Well-Posed Linear System with skew-adjoint generator

Ghislain Haine

ISAE – Supported by IDEX-"Nouveaux Entrants"

Workshop New trends in modeling, control and inverse problems

June, 16–19

Session "Time optimal control and observers"

- 1 Introduction
- 2 The reconstruction algorithm
- 3 Main result
 - With bounded observation operator
 - With unbounded observation operator
- 4 Application
- 5 Conclusion

1 Introduction

2 The reconstruction algorithm

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For instance:

$$A = \begin{bmatrix} 0 & I \\ \Delta & 0 \end{bmatrix} \text{ (+ Dirichlet boundary conditions) on } \Omega \subset \mathbb{R}^n \text{ and}$$

$$X = H_0^1(\Omega) \times L^2(\Omega).$$

↓

the classical wave equation.

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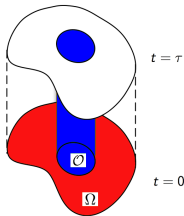
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The classical wave equation, with $C = \begin{bmatrix} 0 & \chi_O \end{bmatrix}$:

$$\begin{aligned} y(t) &= \begin{bmatrix} 0 & \chi_O \end{bmatrix} \begin{bmatrix} w(t) \\ \dot{w}(t) \end{bmatrix}, & \forall t \in [0, \tau], \\ &= \chi_O \dot{w}(t), & \forall t \in [0, \tau]. \end{aligned}$$



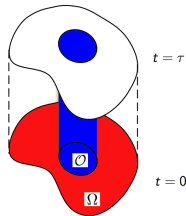
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Our problem

Reconstruct the unknown z_0 from the measurement $y(t)$.

1 Introduction

2 The reconstruction algorithm

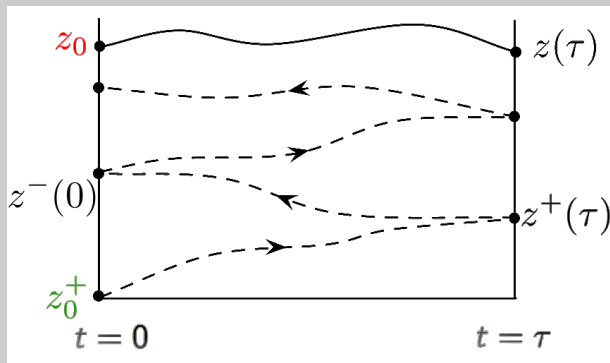
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Intuitive representation



2 iterations, observation on $[0, \tau]$.

Some bibliography

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- **2008:** Phung and Zhang (*SIAM J. Appl. Math.*) introduced the Time Reversal Focusing (TRF), for the Kirchhoff plate equation
- **2010:** Ramdani, Tucsnak and Weiss (*Automatica*) generalized the TRF, based on the generalization of Luenberger's observers

We construct the **forward observer**

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which is known to be exponentially stable if and only if (A, C) is exactly observable, *i.e.*

$$\exists \tau > 0, \exists k_\tau > 0, \int_0^\tau \|y(t)\|^2 dt \geq k_\tau^2 \|z_0\|^2, \quad \forall z_0 \in \mathcal{D}(A).$$

Exponential stability $\Rightarrow \exists M > 0, \beta > 0$ such that

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$$\begin{cases} \dot{z}^-(t) = Az^-(t) + C^*Cz^-(t) - C^*y(t), & \forall t \in [0, \tau], \\ z^-(\tau) = z^+(\tau). \end{cases}$$

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After a time reversal $Z^-(t) = \mathfrak{A}_\tau z^-(t) := z^-(\tau - t)$, we get

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And from similar computations for $A^- := -A - C^*C$ as those for $A^+ := A - C^*C$:

$$t\|z^-(0) - z_0\| \leq M e^{-\beta\tau} \|z^+(\tau) - z(\tau)\| \leq M^2 e^{-2\beta\tau} \|z_0^+ - z_0\|.$$

If the system is exactly observable in time $\tau > 0$, that is if:

$$\exists k_\tau > 0, \int_0^\tau \|y(t)\|^2 dt \geq k_\tau^2 \|z_0\|^2, \quad \forall z_0 \in \mathcal{D}(A),$$

Ito, Ramdani and Tucsnak (Discrete Contin. Dyn. Syst. Ser. S, 2011) proved that

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Iterating n -times the forward-backward observers with $z_n^+(0) = z_{n-1}^-(0)$ leads to

$$\|z_n^-(0) - z_0\| \leq \alpha^n \|z_0^+ - z_0\|.$$

This is the iterative algorithm of Ramdani, Tucsnak and Weiss to reconstruct z_0 from $y(t)$.

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In this work, the exact observability assumption in time τ

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Questions

- Given arbitrary C and $\tau > 0$, does the algorithm converge ?
- If it does, what is the limit of $z_n^-(0)$ and how is it related to z_0 ?

Decomposition of X :

- Let us denote Ψ_τ the following continuous linear operator

$$\begin{array}{rcl} \Psi_\tau & : & X \longrightarrow L^2([0, \tau], Y), \\ & & \color{red}{z_0} \mapsto \color{blue}{y(t)}. \end{array}$$

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- We decompose $X = \text{Ker } \Psi_\tau \oplus (\text{Ker } \Psi_\tau)^\perp$ and define

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Note that the exact observability assumption is equivalent to Ψ_τ is bounded from below and then $\Rightarrow X = \text{Ran } \Psi_\tau^*$.

Stability of the decomposition under the algorithm:

Let us denote \mathbb{T}^+ (resp. \mathbb{T}^-) the semigroup generated by $A^+ := A - C^*C$ (resp. $A^- := -A - C^*C$) on X .

- Forward–backward observers cycle \Rightarrow operator $\mathbb{T}_\tau^- \mathbb{T}_\tau^+$, i.e.

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- Denote \mathbb{S} the group generated by A , then (since $A = A^+ + C^*C$)

$$\mathbb{S}_\tau z_0 = \mathbb{T}_\tau^+ z_0 + \int_0^\tau \mathbb{T}_{\tau-t}^+ C^* \underbrace{C \mathbb{S}_t z_0}_{\Psi_\tau z_0} dt, \quad \forall z_0 \in X.$$

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The algorithm preserves the decomposition of X !

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 - Duhamel formulas $\implies \|L\|_{\mathcal{L}(V_{\text{Obs}})}$ in term of $\inf_{\|z\|=1, z \in V_{\text{Obs}}} \|\Psi_\tau z\|$.
 - $\text{Ran } \Psi_\tau^* \text{ closed in } X \iff \Psi_\tau \text{ bounded from below on } V_{\text{Obs}}$.

Furthermore, it is easy to prove that:

$$z_0^+ \in V_{\text{Obs}} \implies z_n^-(0) \in V_{\text{Obs}}, \quad \forall n \geq 1.$$

Theorem

Denote by Π the orthogonal projection from X onto V_{Obs} . Then the following statements hold true for all $z_0 \in X$ and $z_0^+ \in V_{\text{Obs}}$:

- ❶ For all $n \geq 1$,

$$\|(I - \Pi)(z_n^-(0) - z_0)\| = \|(I - \Pi)z_0\|.$$

- ❷ The sequence $(\|\Pi(z_n^-(0) - z_0)\|)_{n \geq 1}$ is strictly decreasing and

$$\|\Pi(z_n^-(0) - z_0)\| = \|z_n^-(0) - \Pi z_0\| \xrightarrow{n \rightarrow \infty} 0.$$

- ❸ There exists a constant $\alpha \in (0, 1)$, independent of z_0 and z_0^+ , such that for all $n \geq 1$,

$$\|\Pi(z_n^-(0) - z_0)\| \leq \alpha^n \|z_0^+ - \Pi z_0\|,$$

if and only if $\text{Ran } \Psi_\tau^*$ is closed in X .

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- Well-posed linear system

$$\begin{bmatrix} z(t) \\ y|_{[0,t]} \end{bmatrix} = \Sigma_t \begin{bmatrix} z_0 \\ u|_{[0,t]} \end{bmatrix}, \quad \forall t \geq 0,$$

where $u \in \mathcal{U} := L^2([0, \infty), U)$ and $y \in \mathcal{Y} := L^2([0, \infty), Y)$ are the control and the observation (with U and Y two Hilbert spaces).

What happens if C is unbounded ?

- Main issue $\implies A - C^*C$ has no more meaning (as a generator).
How to close the system ?
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- **Well-posedness** means that for all $t \geq 0$:

$$\Sigma_t = \begin{bmatrix} \mathbb{T}_t & \Phi_t \\ \Psi_t & \mathbb{F}_t \end{bmatrix} \in \mathcal{L}(X \times \mathcal{U}, X \times \mathcal{Y}).$$

M. Tucsnak and G. Weiss

Well-posed systems – The LTI case and beyond (Automatica, 2014)

M. Tucsnak and G. Weiss

Well-posed systems – The LTI case and beyond ([Automatica, 2014](#))

Let $A \in \mathcal{L}(\mathcal{D}(A), X)$ be the infinitesimal generator of \mathbb{T} .

We denote X_1 the Hilbert space $\mathcal{D}(A)$ (with the graph norm) and X_{-1} its dual with respect to the pivot space X .

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Associated triple (A, B, C) : There exist a control operator $B \in \mathcal{L}(U, X_{-1})$ and an observation operator $C \in \mathcal{L}(X_1, Y)$ such that

$$\Phi_t u = \int_0^t \mathbb{T}_{t-s} B u(s) ds, \quad \forall u \in \mathcal{U},$$

and

$$\Psi_t z_0(s) = \begin{cases} C \mathbb{T}_s z_0, & \forall s \in [0, t] \\ 0, & \forall s > t \end{cases} \quad \forall z_0 \in X_1.$$

Let Σ be associated with (A, C^*, C) , with A skew-adjoint.

Theorem (Curtain and Weiss 2006)

There exists $\kappa \in (0, \infty]$ such that for all $\gamma \in (0, \kappa)$, the feedback law $u = -\gamma y + v$ (v is the new control) leads to a closed-loop system Σ^γ which is well-posed. Furthermore:

$$\Sigma^\gamma - \Sigma = \Sigma \begin{bmatrix} 0 & 0 \\ 0 & \gamma I \end{bmatrix} \Sigma^\gamma = \Sigma^\gamma \begin{bmatrix} 0 & 0 \\ 0 & \gamma I \end{bmatrix} \Sigma.$$

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Let z^+ be the trajectory of Σ^+ with control $v = \gamma y$ (for simplicity we suppose $u \equiv 0$), then we have

$$z^+(t) - z(t) = \mathbb{T}_t^+ (z_0^+ - z_0), \quad \forall t \geq 0, z_0^+ \in X,$$

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Under some additional assumptions (namely optimizability and estimatability), the closed-loop system is exponentially stable. In other words, the associated semigroup is: z^+ is a **forward observer** of z .

The idea is now to construct the backward observer. There is mainly two ways to do that using the dual of a well-posed linear system.

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Dual system

Define Σ^d by

$$\Sigma_t^d = \begin{bmatrix} \mathbb{T}_t^d & \Phi_t^d \\ \Psi_t^d & \mathbb{F}_t^d \end{bmatrix} = \begin{bmatrix} I & 0 \\ 0 & \mathfrak{R}_t \end{bmatrix} \begin{bmatrix} \mathbb{T}_t^* & \Psi_t^* \\ \Phi_t^* & \mathbb{F}_t^* \end{bmatrix} \begin{bmatrix} I & 0 \\ 0 & \mathfrak{R}_t \end{bmatrix}.$$

Then Σ^d is a well-posed linear system with input space Y , state space X and output space U , associated with (A^*, C^*, B^*) .

Where $\mathfrak{R}_t u(s) := u(t - s)$ is the time reversal operator.

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We then obtain the same theorem as for bounded C , using z^+ and z^- , the respective trajectories of Σ^+ and Σ^- , as forward and backward observers.

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Example

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Consider the following wave system

$$\begin{cases} \ddot{w}(x, t) - \Delta w(x, t) = 0, & \forall x \in \Omega, t > 0, \\ w(x, t) = 0, & \forall x \in \Gamma_0, t > 0, \\ w(x, t) = u(x, t), & \forall x \in \Gamma_1, t > 0, \\ w(x, 0) = w_0(x), \dot{w}(x, 0) = w_1(x), & \forall x \in \Omega, \end{cases}$$

with u the control, and (w_0, w_1) the initial state.

Observation

Let ν be the unit normal vector of Γ_1 pointing towards the exterior of Ω , we observe the system *via*

$$y(x, t) = -\frac{\partial(-\Delta)^{-1}\dot{w}(x, t)}{\partial\nu}, \quad \forall x \in \Gamma_1, t > 0.$$

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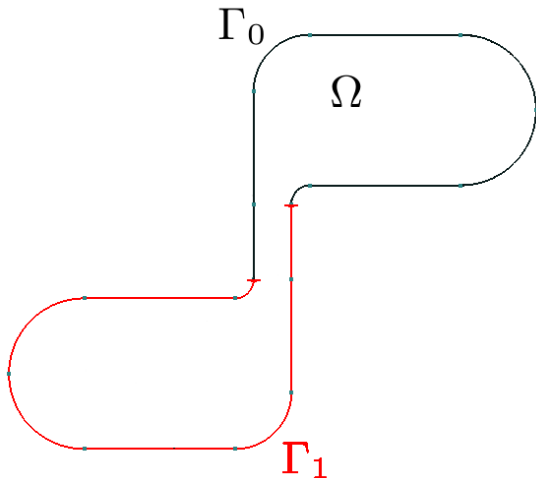
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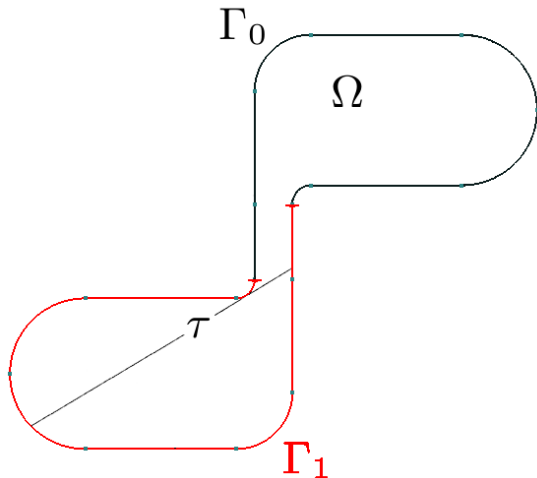
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- So we can use the algorithm.

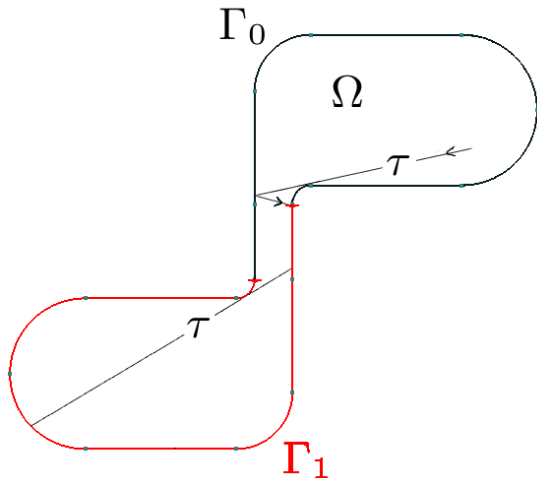
For instance, let us consider the following configuration



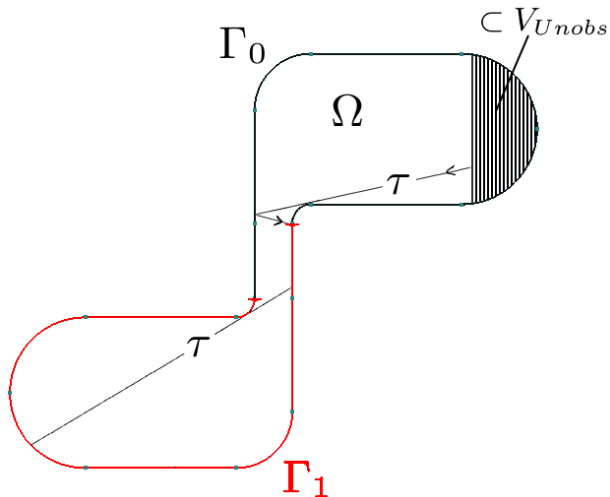
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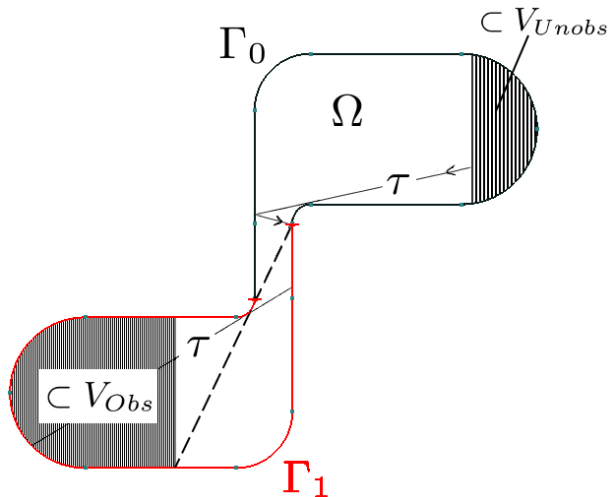
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Choosing a suitable initial data

- $\text{Supp}(w_0)$ has three components W_1, W_2 and W_3 , such that
 - $W_1 \subset V_{\text{Obs}}$
 - $W_2 \subset V_{\text{Unobs}}$
 - $W_3 \cap V_{\text{Obs}} \neq \emptyset$ and $W_3 \cap V_{\text{Unobs}} \neq \emptyset$
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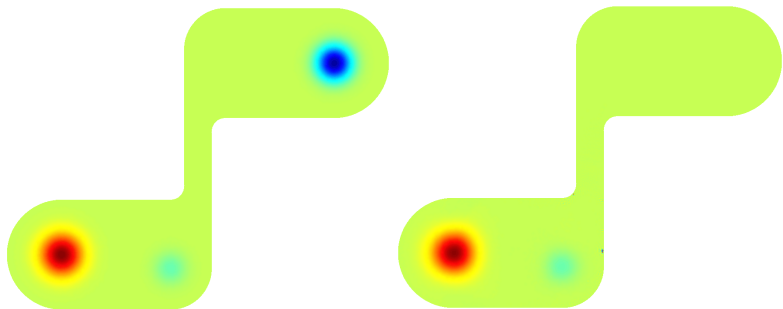
To perform the test, we use

- Gmsh: a 3D finite element grid generator
- GetDP: a general finite element solver

G. Haine and K. Ramdani

Reconstructing initial data using observers: error analysis of the semi-discrete and fully discrete approximations

(Numerische Mathematik (Numer. Math.), 2012)



The initial position (Left) and its reconstruction (Right) after 3 iterations

\Rightarrow 6% of relative error in $L^2(\Omega)$ on the “observable part”.

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More ?

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Still to be done:

- Stability of V_{Obs} and V_{Unobs} with noisy observation y
- Generalization ($A^* \neq -A$)
- Optimization of γ