

Reduced order models for transport

Internship presentation

Master 1 Calcul Scientifique et Mathématiques de l'Information

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Objectives

- 1) Test reduced order model for Burgers' equation.
- 2) Understand reduced order models for linear Hamiltonian systems.
- 3) Test them on piano string model.
- 4) Think about how to improve reduction in the non-linear case.



Reduced order models

We consider the problem

$$\begin{cases} \partial_t u = \mathcal{F}_g(u, \partial_z u, \partial_{zz} u, \dots) \\ u(z, 0) = u_0(z) \end{cases} \quad \begin{matrix} \forall z \in \Omega & \forall t \in I, \\ \forall x \in \Omega. \end{matrix} \quad (1)$$

Hypothesis : It exists a differential manifold M and a *decoding* operator $D : \mathbb{R}^k \rightarrow M$, such that $u(z, t, g) = D(\tilde{u}(z, t, g))$, where $\tilde{u}(z, t, g) \in \mathbb{R}^k$ and $u(\cdot, \cdot, g) \in M$ the solution of (1).

The chain rule gives

$$\dot{u}(z, t, g) = \mathcal{J}_D(\tilde{u}(z, t, g))\dot{\tilde{u}}(z, t, g).$$

Using Galerkin projection, we get

$$\dot{\tilde{u}}(z, t, g) = \mathcal{J}_D^+(\tilde{u}(z, t, g))D(\tilde{u}(z, t, g)),$$

with \mathcal{J}_D^+ the Moore-Penrose inverse of the jacobian of D \mathcal{J}_D .

Symplectic structures [Arnold, 1989]

Definition (Alternate form)

A k -form ω on \mathbb{R}^n is a k -linear skew-symmetric application of \mathbb{R}^n in \mathbb{R} .

Definition (Differential form)

Let M be a differential manifold. A k -differential form ω is a collection of k -forms $\omega_x : (T_x M)^k \rightarrow \mathbb{R}$ defined at each $x \in M$ and varying differentiably with x .

Definition (Symplectic form)

We call a *symplectic form* on M a differential 2-form which is closed and non-degenerate.

The manifold M endowed with this form is called a *symplectic manifold*.

Definition (Symplectic map)

Let (M, ω) and (N, η) be two symplectic manifolds. A differentiable map $f : M \rightarrow N$ is said to be *symplectic* if $f^* \eta = \omega$.

Take $M = \mathbb{R}^{2n}$ and note $x \in \mathbb{R}^{2n}$ as (p, q) with $p, q \in \mathbb{R}^n$. Consider the symplectic form $\omega_{2n}^2 = \sum_{i=1}^n p_i dq_i$ and the usual euclidian structure on \mathbb{R}^{2n} .

We have

$$\omega_{2n}^2(\cdot, \cdot) = \langle \cdot, \mathbb{J}_{2n} \cdot \rangle, \quad \mathbb{J}_{2n} = \begin{pmatrix} 0 & I_n \\ -I_n & 0 \end{pmatrix},$$

so

$$\omega_{2n}^2(A \cdot, A \cdot) = \omega_{2k}^2(\cdot, \cdot) \iff A^T \mathbb{J}_{2n} A = \mathbb{J}_{2k}. \quad (2)$$

Matrices verifying (2) are said to be *symplectic*.

We call $A^+ := \mathbb{J}_{2k} A \mathbb{J}_{2n}$ the *symplectic inverse* of A .

Hamiltonian systems [Arnold, 1989]

Let (M, ω) be a symplectic manifold of even dimension $2n$.

Consider a differentiable function $H : M \rightarrow \mathbb{R}$.

$$\text{Non-degeneracy of } \omega \implies \exists X_H \in \Gamma(M) \mid \omega(X_H, \cdot) = dH.$$

Definition

A *Hamiltonian equation* is a differential equation of the form

$$\dot{z}(t) = X_H(z(t)), \quad \forall t \in I. \quad (3)$$

Proposition

- 1) Solutions of (3) conserve H ,
- 2) ω and the phase space volume are preserved by the phase flow of (3).

Reduced order models for Hamiltonian systems

[Afkham and Hesthaven, 2017]

Goal: for $M = \mathbb{R}^{2n}$, build linear decoder A and encoder B such that :

- A preserves ω^2 ,
- $\mathcal{L}(A) = \|Z - BAZ\|$ has a small value.

Problem: the set of symplectic matrices is not bounded.



Greedy algorithm [Afkham and Hesthaven, 2017]

After iteration k , we have $A_k \in \mathcal{M}_{2n, 2k}$. At iteration $k + 1$:

- 1) Choice of the parameter g_{k+1} maximising

$$\Delta H_k(t) := |H(z(t)) - H(A_k y_k(t))|.$$

- 2) Computation of the solution for this parameter: $S := \{z(t_i, g_{k+1})\}_{i=1, \dots, m}$.

- 3) Choice of the sample with the "worst" projection:

$$s_{k+1} = \operatorname{argmax}_{s \in S} \|s - A_k A_k^+ s\|_2.$$

- 4) Obtain \tilde{v} with Gramm-Schmidt symplectic orthogonalization procedure.

- 5) Set $A_{k+1} = \begin{pmatrix} A_k[:, : k] & | & \tilde{v} & | & A_k[:, : k] & | & \mathbb{J}_{2n} \tilde{v} \end{pmatrix}$.

Application to piano vibrating strings [Chabassier and Joly, 2010]

General model:

$$\partial_{tt}^2 U(x, t) = \partial_x \left[\nabla V(\partial_x U(x, t)) \right], \quad \forall (x, t) \in [0, 1] \times \mathbb{R}_+,$$

$U(x, t) = (v(x, t), u(x, t))$: variations of x on the oscillation plane

V : real differentiable function of \mathbb{R}^{2n} .

with Hamiltonian formulation with:

$$H : (p, q, t) \mapsto \int_{[0,1]} \frac{1}{2} |p|^2 + V(\partial_x q) dx.$$

Linear model: $V(u, v) = \frac{(1-\alpha)}{2} u^2 + \frac{1}{2} v^2$, $\alpha \in [0, 1]$.

Non-linear model: $V(u, v) = \frac{1-\alpha}{2} u^2 + \frac{1}{2} v^2 + \frac{\alpha}{2} (u^2 v + \frac{1}{4} u^4)$, $\alpha \in [0, 1]$.

Application to piano vibrating strings : linear model

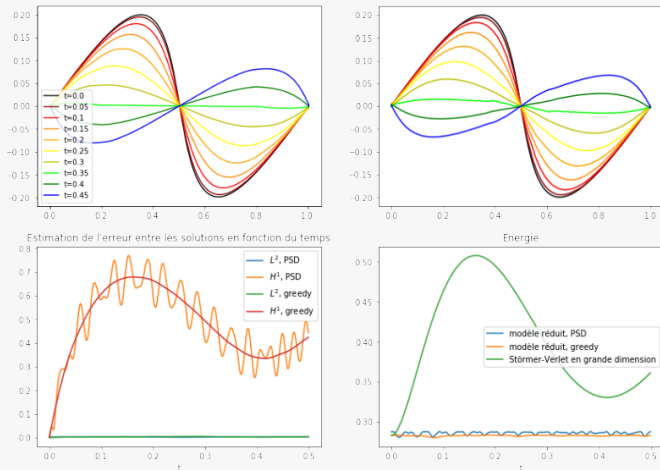


Figure 1: Numerical solution of piano string equation with $V = \frac{1}{2}((1 - g)u^2 + v^2)$, where $g = 0.537$. The reductions were made using PSD with $k = 5$.

Application to piano vibrating strings : non-linear model

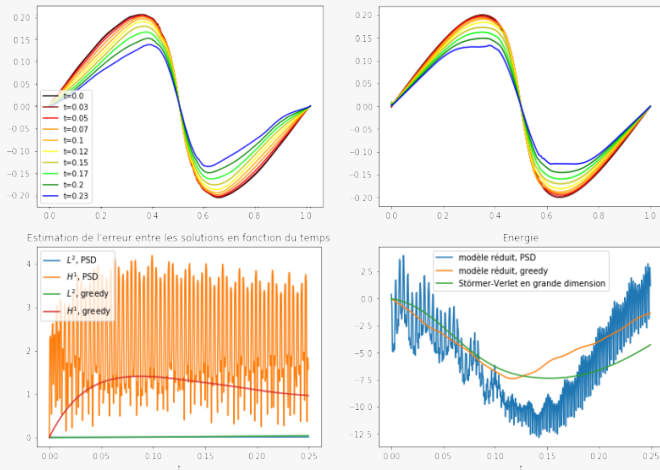


Figure 2: Numerical solution of piano string equation with non-linear model, where $g = 0.8$. The reductions were made using PSD with $k = 5$.

Application to piano vibrating strings : non-linear model

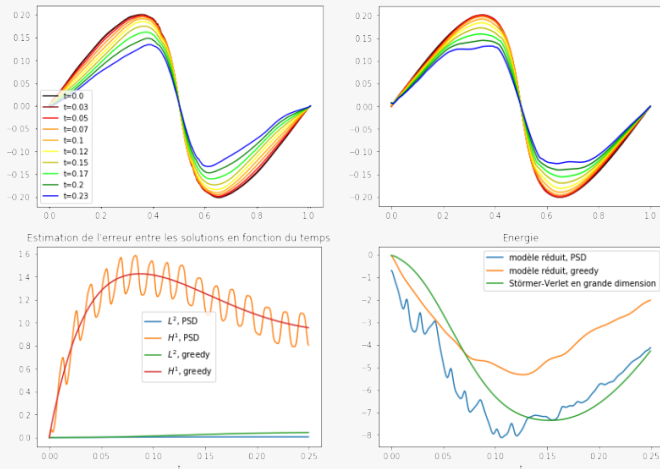


Figure 3: Numerical solution of piano string equation with non-linear model, where $g = 0.8$. The reductions were made using PSD with $k = 20$.

Hyper-reduction

Goal: faster the computation of the reduced model

Kernel regression: build \bar{H} of the form

$$\bar{H}(x) = \sum_{i=1}^m \theta_i K(x^i, x),$$

with K a kernel function.

Loss:

$$\mathcal{L}_{\alpha, \beta}(\theta) = \alpha \sum_{i=1}^m \left\| \frac{A^+ z^{i+1} - A^+ z^i}{\Delta t} - \mathbb{J}_{2k} \nabla \bar{H}(x^i) \right\|^2 + \beta \sum_{i=1}^m |\bar{H}(x^i) - H(Ax^i)|^2.$$

Hyper-reduction : results

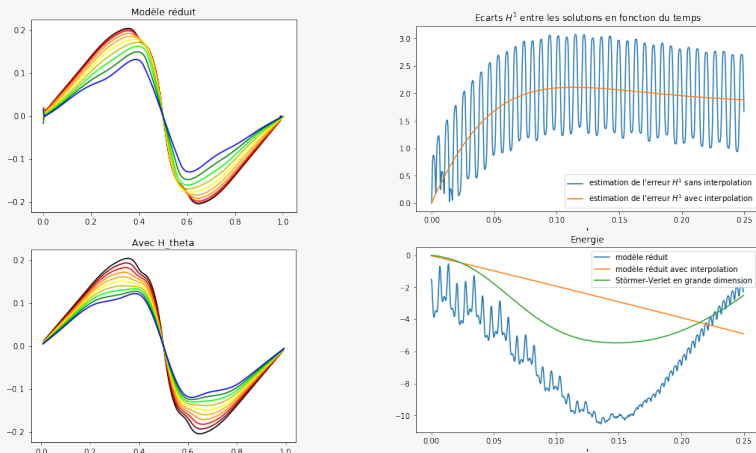


Figure 4: Numerical solution of piano string equation with non-linear model, where $g = 0.783$. The first column represents the piano string position in the oscillation plane for different times computed in low dimension with $H \circ A$ (top) and H_θ (bottom). The second column represents an estimation of the errors made on the string position through time on H^1 norm (top) and the variation of energy during time for the three different solutions (bottom). The reduction was made using PSD with $k = 5$.

Conclusion

- Symplectic structure induce particular constraint on the reduced model,
- useful to find good reduced model
- PSD and greedy algorithm good for linear systems...
- ...but fail for non-linear ones.
- Hyper-reduction gives bad results
- and shows that we have to take into account ∇H (take $\alpha \neq 0$).
- Ideas to look into with generating functions but theoretical difficulties...



Afkham, B. and Hesthaven, J. (2017).

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Preprint at <https://arxiv.org/abs/1703.08345>.



Arnold, V. (1989).

***Mathematical Methods of Classical Mechanics*, chapter 7, 8, 9, 10, pages 163–200.**

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