

On the quadratic finite element approximation of  $1 - d$  waves:  
propagation, observation, control and numerical implementation

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## The continuous model

The 1 –  $d$  wave equation with non-homogeneous boundary conditions:

$$\begin{cases} y_{tt}(x, t) - y_{xx}(x, t) = 0, & x \in (0, 1), t > 0, \\ y(0, t) = 0, y(1, t) = v(t), & t > 0, \\ y(x, 0) = y^0(x), y_t(x, 0) = y^1(x), & x \in (0, 1). \end{cases} \quad (1)$$

**Exact controllability:**  $\forall (y^1, y^0) \in \mathcal{V} = H^{-1} \times L^2, \exists v \in L^2(0, T)$  s.t.  $y(x, T) = y_t(x, T) = 0$ .

The adjoint 1 –  $d$  wave equation with homogeneous boundary conditions:

$$\begin{cases} u_{tt}(x, t) - u_{xx}(x, t) = 0, & x \in (0, 1), t > 0, \\ u(0, t) = u(1, t) = 0, & t > 0, \\ u(x, T) = u^0(x), u_t(x, T) = u^1(x), & x \in (0, 1). \end{cases} \quad (2)$$

Well-posed in the energy space  $\mathcal{V} := H_0^1 \times L^2$ . The energy is conserved in time:

$$\mathcal{E}(u^0, u^1) = \frac{1}{2} (\|u(\cdot, t)\|_{H_0^1}^2 + \|u_t(\cdot, t)\|_{L^2}^2) = \frac{1}{2} (\|u^0\|_{H_0^1}^2 + \|u^1\|_{L^2}^2).$$

The **observability inequality** holds for all solutions of (2), provided  $T \geq 2$ :

$$\mathcal{E}(u^0, u^1) \leq C(T) \int_0^T |u_x(1, t)|^2 dt. \quad (3)$$

**Hilbert Uniqueness Method (HUM):** exact controllability  $\Leftrightarrow$  observability inequality



Lions J.L., *Contrôlabilité exacte, perturbations et stabilisation des systèmes distribués*, Masson, 1988.

# The HUM control

The **HUM control**  $v$  has the explicit form

$$v(t) = \tilde{v}(t) := \tilde{u}_x(1, t), \quad (4)$$

where  $\tilde{u}(x, t)$  is the solution of (2) corresponding to the minimum  $(\tilde{u}^0, \tilde{u}^1) \in \mathcal{V}$  of

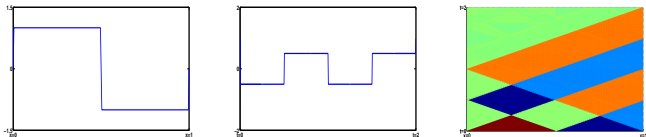
$$j(u^0, u^1) = \frac{1}{2} \int_0^T |u_x(1, t)|^2 dt - \langle (y^1, -y^0), (u(\cdot, 0), u_t(\cdot, 0)) \rangle_{\mathcal{V}', \mathcal{V}}. \quad (5)$$

**Example:**  $y^1 \equiv 0$  and the initial position  $y^0$  given by

$$y^0(x) = H(x) := \begin{cases} 1, & x \in [0, 1/2) \\ -1, & x \in [1/2, 1], \end{cases} \quad (6)$$

for which the optimal control is

$$\tilde{v}(t) = \tilde{v}_H(t) = \begin{cases} -1/2, & t \in (0, 1/2] \cup (1, 3/2], \\ 1/2, & t \in (1/2, 1] \cup (3/2, 2). \end{cases} \quad (7)$$



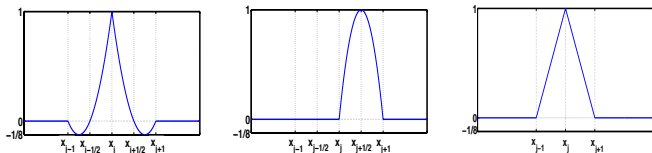
**Figure:** The initial position  $H(x)$  (left) versus the HUM control  $\tilde{v}_H$  (middle) versus the solution of the controlled problem (right) (red = 1, orange = 1/2, green = 0, cyan = -1/2 and blue = -1).

## Linear and quadratic FEM spaces

$N \in \mathbb{N}$ ,  $h = 1/(N + 1)$ . An **uniform grid** of  $[0, 1]$ : **nodes**  $x_j$  and **midpoints**  $x_{j+1/2}$ .

$$\mathcal{U}_1^h := \{u \in H_0^1(0, 1) \text{ s.t. } u|_{I_j} \in \mathcal{P}_1(I_j), 0 \leq j \leq N\} = \text{span}\{\phi_{1,j}^h\}$$

$$\mathcal{U}_2^h := \{u \in H_0^1(0, 1) \text{ s.t. } u|_{I_j} \in \mathcal{P}_2(I_j), 0 \leq j \leq N\} = \text{span}\{\phi_{2,j}^h\} \oplus \text{span}\{\phi_{2,j+1/2}^h\}$$



**Figure:** The basis functions:  $\phi_{2,j}^h$  (left),  $\phi_{2,j+1/2}^h$  (middle) and  $\phi_{1,j}^h$  (right).

Linear/quadratic semi-discretization of the adjoint problem (2):

$$\begin{cases} \text{Find } u_p^h(\cdot, t) \in \mathcal{U}_p^h \text{ s.t. } \frac{d^2}{dt^2}(u_p^h(\cdot, t), \varphi_p^h)_{L^2} + (u_p^h(\cdot, t), \varphi_p^h)_{H_0^1} = 0, \forall \varphi_p^h \in \mathcal{U}_p^h, \\ u_p^h(x, T) = u_p^{h,0}(x), u_{p,t}^h(x, T) = u_p^{h,1}(x), x \in (0, 1), \end{cases} \quad (8)$$

which can be written as a system of second-order linear differential equations (ODEs):

$$M_p^h \mathbf{U}_p^h(t) + S_p^h \mathbf{U}_p^h(t) = 0, \quad \mathbf{U}_p^h(T) = \mathbf{U}_p^{h,0}, \quad \mathbf{U}_{p,t}^h(T) = \mathbf{U}_p^{h,1}, \quad p = 1, 2, \quad (9)$$

where  $M_1^h, S_1^h$  - tri-diagonal and  $M_2^h, S_2^h$  - penta-diagonal mass and stiffness matrices.

## Discrete functional setting

Discrete analogues of  $H_0^1(0, 1)$ ,  $L^2(0, 1)$  or  $H^{-1}(0, 1)$ :

$$\mathcal{H}_p^{h,i} := \{\mathbf{F}_p^h = (F_{p,j/p})_{1 \leq j \leq pN+p-1} \in \mathbb{C}^{pN+p-1} \text{ s.t. } \|\mathbf{F}_p^h\|_{h,i,p} < \infty\}, \quad i = 1, 0, -1.$$

**Inner products** defining the discrete spaces  $\mathcal{H}_p^{h,i}$ ,  $i = 1, 0, -1$ :

$$(\mathbf{E}_p^h, \mathbf{F}_p^h)_{h,i,p} := ((M_p^h(S_p^h)^{-1})^{1-i} S_p^h \mathbf{E}_p^h, \mathbf{F}_p^h)_{p,e}, \quad i = 1, 0, -1.$$

**Discrete energy space** and its **dual**:  $\mathcal{V}_p^h := \mathcal{H}_p^{h,1} \times \mathcal{H}_p^{h,0}$  and  $\mathcal{V}_p^{h,\prime} := \mathcal{H}_p^{h,-1} \times \mathcal{H}_p^{h,0}$ .

Problem (9) is well-posed in  $\mathcal{V}_p^h$ . The total energy is conserved in time:

$$\mathcal{E}_p^h(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = \frac{1}{2} (\|\mathbf{U}_p^h(t)\|_{h,1,p}^2 + \|\mathbf{U}_{p,t}^h(t)\|_{h,0,p}^2) = \frac{1}{2} (\|\mathbf{U}_p^{h,0}\|_{h,1,p}^2 + \|\mathbf{U}_p^{h,1}\|_{h,0,p}^2). \quad (10)$$

**Discrete observability inequality**

$$\mathcal{E}_p^h(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) \leq C_p^h(T) \int_0^T \|B_p^h \mathbf{U}_p^h(t)\|_{p,e}^2 dt, \quad (11)$$

where

$$B_{p,ij} := \begin{cases} -\frac{1}{h}, & (i,j) = (pN+p-1, pN), \\ 0, & \text{otherwise.} \end{cases} \quad (12)$$

The solution of (9) admits the following Fourier representation

$$\mathbf{U}_p^h(t) = \sum_{\pm} \sum_{k=1}^{pN+p-1} \hat{u}_{p,\pm}^k \exp(\pm it \lambda_p^{h,k}) \varphi_p^{h,k}.$$

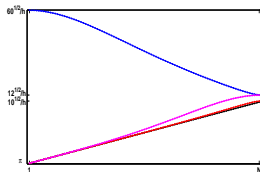


Figure: The square roots of the eigenvalues,  $\lambda_p^h$ : the continuous, acoustic, optic, resonant modes and  $p = 1$ .

$$\lambda_1(\eta) \sim \eta + \eta^3/24 + \eta^5/1920, \quad \lambda_2^a(\eta) \sim \eta + \eta^5/1440.$$

$C_p^h(T)$  in (11) blows-up. !!!



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Micu S., *Uniform boundary controllability of a semi-discrete 1-D wave equation*, Numer. Math., 2002.

## Bi-grid filtering algorithm. Uniform observability results.

$$\mathcal{B}_1^h := \{F_1^h = (F_{1,j})_{1 \leq j \leq N} \text{ s.t. } F_{1,2j+1} = (F_{1,2j} + F_{1,2j+2})/2, \forall 0 \leq j \leq (N-1)/2\}$$

$$\mathcal{B}_2^h := \{F_2^h = (F_{2,j/2})_{1 \leq j \leq 2N+1} \text{ s.t. } F_{2,j+1/2} = (F_{2,j} + F_{2,j+1})/2, \forall 0 \leq j \leq N, \\ \text{and } F_{2,2j+1} = (F_{2,2j} + F_{2,2j+2})/2, \forall 0 \leq j \leq (N-1)/2\}.$$

### Theorem

For all initial data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) \in (\mathcal{B}_p^h \times \mathcal{B}_p^h) \cap \mathcal{V}_p^h$  in the adjoint problem (9) and for all  $T \geq 2$ , the observability inequality (11) holds uniformly as  $h \rightarrow 0$ .

Uniform observability results for the fully discrete conservative scheme:

$$\mathbf{U}_p^{h,k+1} - 2\mathbf{U}_p^{h,k} + \mathbf{U}_p^{h,k-1} + (\delta t)^2 (M^h)^{-1} S^h (\mathbf{U}_p^{h,k+1} + \mathbf{U}_p^{h,k-1})/2 = 0.$$



Ervedoza S., Zheng C., Zuazua E., *On the observability of time-discrete conservative linear systems*, J. Functional Analysis, 254(12)(2008), 3037–3078.



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Loreti P., Mehrenberger M., *An Ingham type proof for a two-grid observability theorem*, EAIM: COCV, 2008.



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Negreanu M., Zuazua E., *Convergence of a multigrid method for the controllability of a 1-d wave*



For a particular solution  $\tilde{\mathbf{U}}_p^h(t)$  of the adjoint problem (9), consider the non-homogeneous problem

$$M_p^h \mathbf{Y}_{p,tt}^h(t) + S_p^h \mathbf{Y}_p^h(t) = -(B_p^h)^* B_p^h \tilde{\mathbf{U}}_p^h(t), \quad \mathbf{Y}_p^h(0) = \mathbf{Y}_p^{h,0}, \quad \mathbf{Y}_{p,t}^h(0) = \mathbf{Y}_p^{h,1}. \quad (13)$$

The discrete quadratic functional:

$$\mathcal{J}_p^h(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = \frac{1}{2} \int_0^T \|B_p^h \mathbf{U}_p^h(t)\|_{p,e}^2 dt - \langle (\mathbf{Y}_p^{h,1}, -\mathbf{Y}_p^{h,0}), (\mathbf{U}_p^h(0), \mathbf{U}_{p,t}^h(0)) \rangle_{\mathcal{V}_p^{h,t}, \mathcal{V}_p^h}, \quad (14)$$

$\mathbf{U}_p^h(t)$  being the solution of the adjoint problem (9) with initial data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1})$ .

The uniform observability inequality (11) in the class of initial data  $\mathcal{B}_p^h \times \mathcal{B}_h^p$  implies

- the **coercivity** of  $\mathcal{J}_p^h$
- the **convergence** of the last component of  $B_p^h \tilde{\mathbf{U}}_p^h(t)$  to the continuous optimal control  $\tilde{v}$ .



$\tilde{\mathbf{H}}_p^h := (\tilde{H}_{p,j/p})_{1 \leq j \leq pN+p-1}$ , where  $\tilde{H}_{p,j/p} = (H, \phi_{p,j/p}^h)_{L^2}$ , for all  $1 \leq j \leq pN+p-1$ .

The numerical approximation of  $H(x)$  we consider is  $\mathbf{Y}_p^{h,0} = \mathbf{H}_p^h := (M_p^h)^{-1} \tilde{\mathbf{H}}_p^h$ .

Some projections:

$$\mathbf{H}_{1,lo}^h = \sum_{k=1}^{(N-1)/2} (\mathbf{H}_1^h, \varphi_1^{h,k})_{h,0,1} \varphi_1^{h,k}, \quad \mathbf{H}_{1,hi}^h = \sum_{k=(N+1)/2}^N (\mathbf{H}_1^h, \varphi_1^{h,k})_{h,0,1} \varphi_1^{h,k},$$

$$\mathbf{H}_{2,lo}^{h,\alpha} = \sum_{k=1}^{(N-1)/2} (\mathbf{H}_2^h, \varphi_2^{h,\alpha,k})_{h,0,2} \varphi_2^{h,\alpha,k}, \quad \mathbf{H}_{2,hi}^{h,\alpha} = \sum_{k=(N+1)/2}^N (\mathbf{H}_2^h, \varphi_2^{h,\alpha,k})_{h,0,2} \varphi_2^{h,\alpha,k}, \quad \alpha \in \{a, o\}.$$

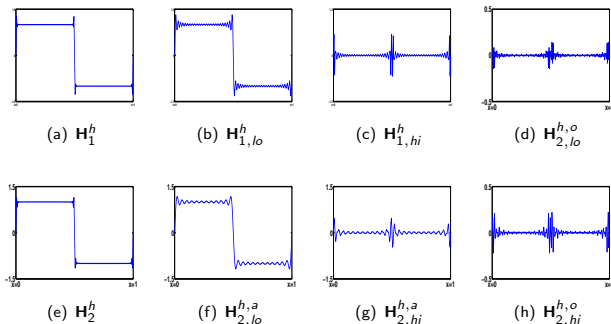


Figure: The discrete Heaviside functions  $\mathbf{H}_p^h$  and their projections.

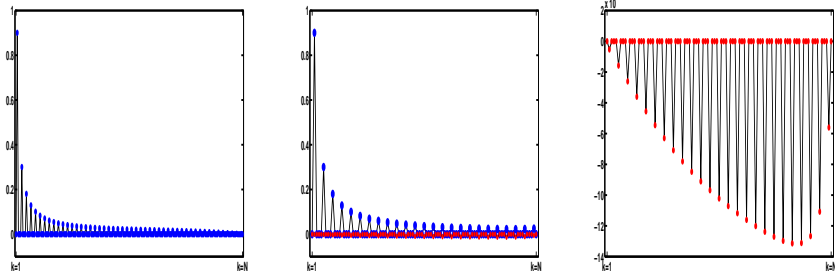


Figure: The Fourier coefficients of  $\mathbf{H}_p^h$  for  $p = 1$  (left),  $p = 2$  (center, blue=acoustic, red=optic),  $p = 2$  - the optic branch (right).

## Conjugate gradient algorithm without filtering

**Step 1.** Solve the adjoint problem (9) with arbitrary data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1}) \in \mathcal{V}_p^h$ , for example the trivial one. This step yields the solution  $\mathbf{U}_{p,0}^h(t)$ .

## Conjugate gradient algorithm without filtering

**Step 1.** Solve the adjoint problem (9) with arbitrary data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1}) \in \mathcal{V}_p^h$ , for example the trivial one. This step yields the solution  $\mathbf{U}_{p,0}^h(t)$ .

**Step 2.** Compute the first gradient  $(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1}) := \nabla \mathcal{J}_p^h(\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1})$  by solving (13) with initial data  $(\mathbf{Y}_{p,0}^{h,0}, \mathbf{Y}_{p,0}^{h,1})$  and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{U}_{p,0}^h(t)$ . This produces the solution  $\mathbf{Y}_{p,0}^h(t)$ . Then

$$\mathbf{G}_{p,0}^{h,0} = -(S_p^h)^{-1} M_p^h \mathbf{Y}_{p,0,t}^h(T) \text{ and } \mathbf{G}_{p,0}^{h,1} = \mathbf{Y}_{p,0}^h(T).$$

## Conjugate gradient algorithm without filtering

**Step 1.** Solve the adjoint problem (9) with arbitrary data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1}) \in \mathcal{V}_p^h$ , for example the trivial one. This step yields the solution  $\mathbf{U}_{p,0}^h(t)$ .

**Step 2.** Compute the first gradient  $(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1}) := \nabla \mathcal{J}_p^h(\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1})$  by solving (13) with initial data  $(\mathbf{Y}_{p,0}^{h,0}, \mathbf{Y}_{p,0}^{h,1})$  and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{U}_{p,0}^h(t)$ . This produces the solution  $\mathbf{Y}_{p,0}^h(t)$ . Then  $\mathbf{G}_{p,0}^{h,0} = -(S_p^h)^{-1} M_p^h \mathbf{Y}_{p,0,t}^h(T)$  and  $\mathbf{G}_{p,0}^{h,1} = \mathbf{Y}_{p,0}^h(T)$ .

**Step 3.** If  $(\|\mathbf{G}_{p,0}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,0}^{h,1}\|_{h,0,p}^2)^{1/2} \geq \epsilon$ , compute the first descent direction  $(\mathbf{D}_{p,0}^{h,0}, \mathbf{D}_{p,0}^{h,1}) = -(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1})$ .

## Conjugate gradient algorithm without filtering

**Step 1.** Solve the adjoint problem (9) with arbitrary data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1}) \in \mathcal{V}_p^h$ , for example the trivial one. This step yields the solution  $\mathbf{U}_{p,0}^h(t)$ .

**Step 2.** Compute the first gradient  $(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1}) := \nabla \mathcal{J}_p^h(\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1})$  by solving (13) with initial data  $(\mathbf{Y}_{p,0}^{h,0}, \mathbf{Y}_{p,0}^{h,1})$  and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{U}_{p,0}^h(t)$ . This produces the solution  $\mathbf{Y}_{p,0}^h(t)$ . Then  $\mathbf{G}_{p,0}^{h,0} = -(S_p^h)^{-1} M_p^h \mathbf{Y}_{p,0,t}^h(T)$  and  $\mathbf{G}_{p,0}^{h,1} = \mathbf{Y}_{p,0}^h(T)$ .

**Step 3.** If  $(\|\mathbf{G}_{p,0}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,0}^{h,1}\|_{h,0,p}^2)^{1/2} \geq \epsilon$ , compute the first descent direction  $(\mathbf{D}_{p,0}^{h,0}, \mathbf{D}_{p,0}^{h,1}) = -(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1})$ .

**Step 4.** Given  $(\mathbf{U}_{p,n}^{h,0}, \mathbf{U}_{p,n}^{h,1})$ ,  $(\mathbf{G}_{p,n}^{h,0}, \mathbf{G}_{p,n}^{h,1})$  and  $(\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  in  $\mathcal{V}_p^h$ , compute them  $n + 1$ :

## Conjugate gradient algorithm without filtering

**Step 1.** Solve the adjoint problem (9) with arbitrary data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1}) \in \mathcal{V}_p^h$ , for example the trivial one. This step yields the solution  $\mathbf{U}_{p,0}^h(t)$ .

**Step 2.** Compute the first gradient  $(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1}) := \nabla \mathcal{J}_p^h(\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1})$  by solving (13) with initial data  $(\mathbf{Y}_{p,0}^{h,0}, \mathbf{Y}_{p,0}^{h,1})$  and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{U}_{p,0}^h(t)$ . This produces the solution  $\mathbf{Y}_{p,0}^h(t)$ . Then  $\mathbf{G}_{p,0}^{h,0} = -(S_p^h)^{-1} M_p^h \mathbf{Y}_{p,0,t}^h(T)$  and  $\mathbf{G}_{p,0}^{h,1} = \mathbf{Y}_{p,0}^h(T)$ .

**Step 3.** If  $(\|\mathbf{G}_{p,0}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,0}^{h,1}\|_{h,0,p}^2)^{1/2} \geq \epsilon$ , compute the first descent direction  $(\mathbf{D}_{p,0}^{h,0}, \mathbf{D}_{p,0}^{h,1}) = -(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1})$ .

**Step 4.** Given  $(\mathbf{U}_{p,n}^{h,0}, \mathbf{U}_{p,n}^{h,1})$ ,  $(\mathbf{G}_{p,n}^{h,0}, \mathbf{G}_{p,n}^{h,1})$  and  $(\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  in  $\mathcal{V}_p^h$ , compute them  $n+1$ :

**Step 4.a.** Solve (9) with data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  and denote the solution by  $\mathbf{D}_{p,n}^h(t)$ .

## Conjugate gradient algorithm without filtering

**Step 1.** Solve the adjoint problem (9) with arbitrary data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1}) \in \mathcal{V}_p^h$ , for example the trivial one. This step yields the solution  $\mathbf{U}_{p,0}^h(t)$ .

**Step 2.** Compute the first gradient  $(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1}) := \nabla \mathcal{J}_p^h(\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1})$  by solving (13) with initial data  $(\mathbf{Y}_{p,0}^{h,0}, \mathbf{Y}_{p,0}^{h,1})$  and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{U}_{p,0}^h(t)$ . This produces the solution  $\mathbf{Y}_{p,0}^h(t)$ . Then  $\mathbf{G}_{p,0}^{h,0} = -(S_p^h)^{-1} M_p^h \mathbf{Y}_{p,0,t}^h(T)$  and  $\mathbf{G}_{p,0}^{h,1} = \mathbf{Y}_{p,0}^h(T)$ .

**Step 3.** If  $(\|\mathbf{G}_{p,0}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,0}^{h,1}\|_{h,0,p}^2)^{1/2} \geq \epsilon$ , compute the first descent direction  $(\mathbf{D}_{p,0}^{h,0}, \mathbf{D}_{p,0}^{h,1}) = -(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1})$ .

**Step 4.** Given  $(\mathbf{U}_{p,n}^{h,0}, \mathbf{U}_{p,n}^{h,1})$ ,  $(\mathbf{G}_{p,n}^{h,0}, \mathbf{G}_{p,n}^{h,1})$  and  $(\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  in  $\mathcal{V}_p^h$ , compute them  $n+1$ :

**Step 4.a.** Solve (9) with data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  and denote the solution by  $\mathbf{D}_{p,n}^h(t)$ .

**Step 4.b.** Solve (13) with trivial initial data and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{D}_{p,n}^h(t)$  and denote the solution by  $\mathbf{Y}_{p,n+1}^h(t)$ . Take  $\mathbf{Z}_{p,n}^{h,0} = -(S_p^h)^{-1} M_p^h \mathbf{Y}_{p,n+1,t}^h(T)$  and  $\mathbf{Z}_{p,n}^{h,1} = \mathbf{Y}_{p,n+1}^h(T)$ .



## Conjugate gradient algorithm without filtering

**Step 1.** Solve the adjoint problem (9) with arbitrary data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1}) \in \mathcal{V}_p^h$ , for example the trivial one. This step yields the solution  $\mathbf{U}_{p,0}^h(t)$ .

**Step 2.** Compute the first gradient  $(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1}) := \nabla \mathcal{J}_p^h(\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1})$  by solving (13) with initial data  $(\mathbf{Y}_{p,0}^{h,0}, \mathbf{Y}_{p,0}^{h,1})$  and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{U}_{p,0}^h(t)$ . This produces the solution  $\mathbf{Y}_{p,0}^h(t)$ . Then  $\mathbf{G}_{p,0}^{h,0} = -(S_p^h)^{-1} M_p^h \mathbf{Y}_{p,0,t}^h(T)$  and  $\mathbf{G}_{p,0}^{h,1} = \mathbf{Y}_{p,0}^h(T)$ .

**Step 3.** If  $(\|\mathbf{G}_{p,0}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,0}^{h,1}\|_{h,0,p}^2)^{1/2} \geq \epsilon$ , compute the first descent direction  $(\mathbf{D}_{p,0}^{h,0}, \mathbf{D}_{p,0}^{h,1}) = -(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1})$ .

**Step 4.** Given  $(\mathbf{U}_{p,n}^{h,0}, \mathbf{U}_{p,n}^{h,1})$ ,  $(\mathbf{G}_{p,n}^{h,0}, \mathbf{G}_{p,n}^{h,1})$  and  $(\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  in  $\mathcal{V}_p^h$ , compute them  $n+1$ :

**Step 4.a.** Solve (9) with data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  and denote the solution by  $\mathbf{D}_{p,n}^h(t)$ .

**Step 4.b.** Solve (13) with trivial initial data and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{D}_{p,n}^h(t)$  and denote the solution by  $\mathbf{Y}_{p,n+1}^h(t)$ . Take  $\mathbf{Z}_{p,n}^{h,0} = -(S_p^h)^{-1} M_p^h \mathbf{Y}_{p,n+1,t}^h(T)$  and  $\mathbf{Z}_{p,n}^{h,1} = \mathbf{Y}_{p,n+1}^h(T)$ .

**Step 4.c.** Set  $\rho_{p,n} := -\frac{\|\mathbf{G}_{p,n}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,n}^{h,1}\|_{h,0,p}^2}{(\mathbf{Z}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,0})_{h,1,p} + (\mathbf{Z}_{p,n}^{h,1}, \mathbf{D}_{p,n}^{h,1})_{h,0,p}}$ .

## Conjugate gradient algorithm without filtering

**Step 1.** Solve the adjoint problem (9) with arbitrary data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1}) \in \mathcal{V}_p^h$ , for example the trivial one. This step yields the solution  $\mathbf{U}_{p,0}^h(t)$ .

**Step 2.** Compute the first gradient  $(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1}) := \nabla \mathcal{J}_p^h(\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1})$  by solving (13) with initial data  $(\mathbf{Y}_{p,0}^{h,0}, \mathbf{Y}_{p,0}^{h,1})$  and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{U}_{p,0}^h(t)$ . This produces the solution  $\mathbf{Y}_{p,0}^h(t)$ . Then  $\mathbf{G}_{p,0}^{h,0} = -(\mathcal{S}_p^h)^{-1} M_p^h \mathbf{Y}_{p,0,t}^h(T)$  and  $\mathbf{G}_{p,0}^{h,1} = \mathbf{Y}_{p,0}^h(T)$ .

**Step 3.** If  $(\|\mathbf{G}_{p,0}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,0}^{h,1}\|_{h,0,p}^2)^{1/2} \geq \epsilon$ , compute the first descent direction  $(\mathbf{D}_{p,0}^{h,0}, \mathbf{D}_{p,0}^{h,1}) = -(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1})$ .

**Step 4.** Given  $(\mathbf{U}_{p,n}^{h,0}, \mathbf{U}_{p,n}^{h,1})$ ,  $(\mathbf{G}_{p,n}^{h,0}, \mathbf{G}_{p,n}^{h,1})$  and  $(\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  in  $\mathcal{V}_p^h$ , compute them  $n+1$ :

**Step 4.a.** Solve (9) with data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  and denote the solution by  $\mathbf{D}_{p,n}^h(t)$ .

**Step 4.b.** Solve (13) with trivial initial data and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{D}_{p,n}^h(t)$  and denote the solution by  $\mathbf{Y}_{p,n+1}^h(t)$ . Take  $\mathbf{Z}_{p,n}^{h,0} = -(\mathcal{S}_p^h)^{-1} M_p^h \mathbf{Y}_{p,n+1,t}^h(T)$  and  $\mathbf{Z}_{p,n}^{h,1} = \mathbf{Y}_{p,n+1}^h(T)$ .

**Step 4.c.** Set  $\rho_{p,n} := -\frac{\|\mathbf{G}_{p,n}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,n}^{h,1}\|_{h,0,p}^2}{(\mathbf{Z}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,0})_{h,1,p} + (\mathbf{Z}_{p,n}^{h,1}, \mathbf{D}_{p,n}^{h,1})_{h,0,p}}$ .

**Step 4.d.**  $(\mathbf{U}_{p,n+1}^{h,0}, \mathbf{U}_{p,n+1}^{h,1}) := (\mathbf{U}_{p,n}^{h,0}, \mathbf{U}_{p,n}^{h,1}) + \rho_{p,n}(\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$ .

## Conjugate gradient algorithm without filtering

**Step 1.** Solve the adjoint problem (9) with arbitrary data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1}) \in \mathcal{V}_p^h$ , for example the trivial one. This step yields the solution  $\mathbf{U}_{p,0}^h(t)$ .

**Step 2.** Compute the first gradient  $(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1}) := \nabla \mathcal{J}_p^h(\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1})$  by solving (13) with initial data  $(\mathbf{Y}_{p,0}^{h,0}, \mathbf{Y}_{p,0}^{h,1})$  and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{U}_{p,0}^h(t)$ . This produces the solution  $\mathbf{Y}_{p,0}^h(t)$ . Then  $\mathbf{G}_{p,0}^{h,0} = -(\mathcal{S}_p^h)^{-1} M_p^h \mathbf{Y}_{p,0,t}^h(T)$  and  $\mathbf{G}_{p,0}^{h,1} = \mathbf{Y}_{p,0}^h(T)$ .

**Step 3.** If  $(\|\mathbf{G}_{p,0}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,0}^{h,1}\|_{h,0,p}^2)^{1/2} \geq \epsilon$ , compute the first descent direction  $(\mathbf{D}_{p,0}^{h,0}, \mathbf{D}_{p,0}^{h,1}) = -(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1})$ .

**Step 4.** Given  $(\mathbf{U}_{p,n}^{h,0}, \mathbf{U}_{p,n}^{h,1})$ ,  $(\mathbf{G}_{p,n}^{h,0}, \mathbf{G}_{p,n}^{h,1})$  and  $(\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  in  $\mathcal{V}_p^h$ , compute them  $n+1$ :

**Step 4.a.** Solve (9) with data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  and denote the solution by  $\mathbf{D}_{p,n}^h(t)$ .

**Step 4.b.** Solve (13) with trivial initial data and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{D}_{p,n}^h(t)$  and denote the solution by  $\mathbf{Y}_{p,n+1}^h(t)$ . Take  $\mathbf{Z}_{p,n}^{h,0} = -(\mathcal{S}_p^h)^{-1} M_p^h \mathbf{Y}_{p,n+1,t}^h(T)$  and  $\mathbf{Z}_{p,n}^{h,1} = \mathbf{Y}_{p,n+1}^h(T)$ .

**Step 4.c.** Set  $\rho_{p,n} := -\frac{\|\mathbf{G}_{p,n}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,n}^{h,1}\|_{h,0,p}^2}{(\mathbf{Z}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,0})_{h,1,p} + (\mathbf{Z}_{p,n}^{h,1}, \mathbf{D}_{p,n}^{h,1})_{h,0,p}}$ .

**Step 4.d.**  $(\mathbf{U}_{p,n+1}^{h,0}, \mathbf{U}_{p,n+1}^{h,1}) := (\mathbf{U}_{p,n}^{h,0}, \mathbf{U}_{p,n}^{h,1}) + \rho_{p,n}(\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$ .

**Step 4.e.**  $(\mathbf{G}_{p,n+1}^{h,0}, \mathbf{G}_{p,n+1}^{h,1}) := (\mathbf{G}_{p,n}^{h,0}, \mathbf{G}_{p,n}^{h,1}) + \rho_{p,n}(\mathbf{Z}_{p,n}^{h,0}, \mathbf{Z}_{p,n}^{h,1})$ .

## Conjugate gradient algorithm without filtering

**Step 1.** Solve the adjoint problem (9) with arbitrary data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1}) \in \mathcal{V}_p^h$ , for example the trivial one. This step yields the solution  $\mathbf{U}_{p,0}^h(t)$ .

**Step 2.** Compute the first gradient  $(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1}) := \nabla \mathcal{J}_p^h(\mathbf{U}_{p,0}^{h,0}, \mathbf{U}_{p,0}^{h,1})$  by solving (13) with initial data  $(\mathbf{Y}_{p,0}^{h,0}, \mathbf{Y}_{p,0}^{h,1})$  and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{U}_{p,0}^h(t)$ . This produces the solution  $\mathbf{Y}_{p,0}^h(t)$ . Then  $\mathbf{G}_{p,0}^{h,0} = -(S_p^h)^{-1} M_p^h \mathbf{Y}_{p,0,t}^h(T)$  and  $\mathbf{G}_{p,0}^{h,1} = \mathbf{Y}_{p,0}^h(T)$ .

**Step 3.** If  $(\|\mathbf{G}_{p,0}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,0}^{h,1}\|_{h,0,p}^2)^{1/2} \geq \epsilon$ , compute the first descent direction  $(\mathbf{D}_{p,0}^{h,0}, \mathbf{D}_{p,0}^{h,1}) = -(\mathbf{G}_{p,0}^{h,0}, \mathbf{G}_{p,0}^{h,1})$ .

**Step 4.** Given  $(\mathbf{U}_{p,n}^{h,0}, \mathbf{U}_{p,n}^{h,1})$ ,  $(\mathbf{G}_{p,n}^{h,0}, \mathbf{G}_{p,n}^{h,1})$  and  $(\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  in  $\mathcal{V}_p^h$ , compute them  $n+1$ :

**Step 4.a.** Solve (9) with data  $(\mathbf{U}_p^{h,0}, \mathbf{U}_p^{h,1}) = (\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$  and denote the solution by  $\mathbf{D}_{p,n}^h(t)$ .

**Step 4.b.** Solve (13) with trivial initial data and  $\tilde{\mathbf{U}}_p^h(t) = \mathbf{D}_{p,n}^h(t)$  and denote the solution by  $\mathbf{Y}_{p,n+1}^h(t)$ . Take  $\mathbf{Z}_{p,n}^{h,0} = -(S_p^h)^{-1} M_p^h \mathbf{Y}_{p,n+1,t}^h(T)$  and  $\mathbf{Z}_{p,n}^{h,1} = \mathbf{Y}_{p,n+1}^h(T)$ .

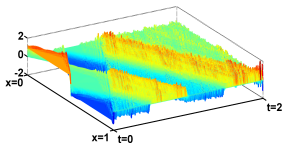
**Step 4.c.** Set  $\rho_{p,n} := -\frac{\|\mathbf{G}_{p,n}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,n}^{h,1}\|_{h,0,p}^2}{(\mathbf{Z}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,0})_{h,1,p} + (\mathbf{Z}_{p,n}^{h,1}, \mathbf{D}_{p,n}^{h,1})_{h,0,p}}$ .

**Step 4.d.**  $(\mathbf{U}_{p,n+1}^{h,0}, \mathbf{U}_{p,n+1}^{h,1}) := (\mathbf{U}_{p,n}^{h,0}, \mathbf{U}_{p,n}^{h,1}) + \rho_{p,n}(\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$ .

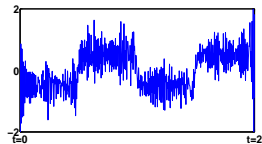
**Step 4.e.**  $(\mathbf{G}_{p,n+1}^{h,0}, \mathbf{G}_{p,n+1}^{h,1}) := (\mathbf{G}_{p,n}^{h,0}, \mathbf{G}_{p,n}^{h,1}) + \rho_{p,n}(\mathbf{Z}_{p,n}^{h,0}, \mathbf{Z}_{p,n}^{h,1})$ .

**Step 4.f.**  $(\mathbf{D}_{p,n+1}^{h,0}, \mathbf{D}_{p,n+1}^{h,1}) := -(\mathbf{G}_{p,n+1}^{h,0}, \mathbf{G}_{p,n+1}^{h,1}) + \frac{\|\mathbf{G}_{p,n+1}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,n+1}^{h,1}\|_{h,0,p}^2}{\|\mathbf{G}_{p,n}^{h,0}\|_{h,1,p}^2 + \|\mathbf{G}_{p,n}^{h,1}\|_{h,0,p}^2} (\mathbf{D}_{p,n}^{h,0}, \mathbf{D}_{p,n}^{h,1})$ .

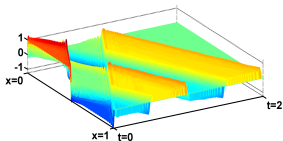
# Numerical results (I)



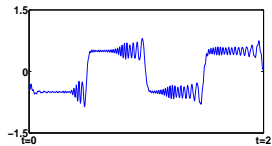
(a) Solution for  $\mathbf{Y}_1^{h,0} = \mathbf{H}_1^h, \mathbf{Y}_1^{h,1} = 0$



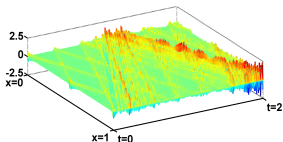
(b) Control for  $\mathbf{Y}_1^{h,0} = \mathbf{H}_1^h, \mathbf{Y}_1^{h,1} = 0$



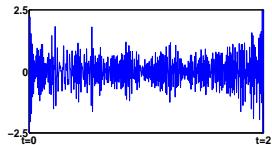
(c) Solution for  $\mathbf{Y}_1^{h,0} = \mathbf{H}_{1,lo}^h, \mathbf{Y}_1^{h,1} = 0$



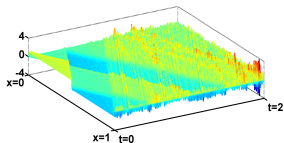
(d) Control for  $\mathbf{Y}_1^{h,0} = \mathbf{H}_{1,lo}^h, \mathbf{Y}_1^{h,1} = 0$



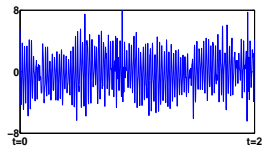
(e) Solution for  $\mathbf{Y}_1^{h,0} = \mathbf{H}_{1,hi}^h, \mathbf{Y}_1^{h,1} = 0$



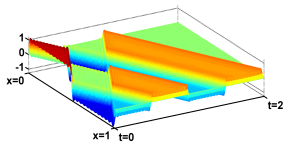
(f) Control for  $\mathbf{Y}_1^{h,0} = \mathbf{H}_{1,hi}^h, \mathbf{Y}_1^{h,1} = 0$



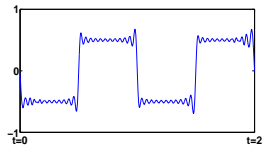
(a) Solution for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_2^h, \mathbf{Y}_2^{h,1} = 0$



(b) Control for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_2^h, \mathbf{Y}_2^{h,1} = 0$

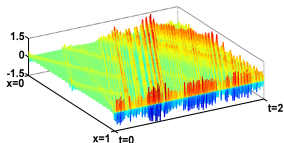


(c) Solution for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_{2,lo}^{h,a}, \mathbf{Y}_2^{h,1} = 0$

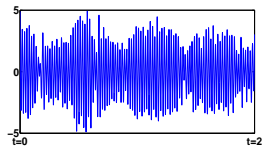


(d) Control for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_{2,lo}^{h,a}, \mathbf{Y}_2^{h,1} = 0$

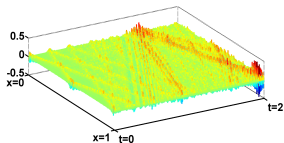
Figure: Solutions of the controlled problem (13) and the corresponding numerical controls for  $p = 2$  arising by minimizing  $\mathcal{J}_2^h$  over the whole space  $\mathcal{V}_2^h$ .



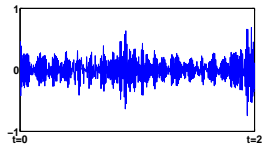
(a) Solution for  $Y_2^{h,0} = H_{2,hi}^{h,a} Y_2^{h,1} = 0$



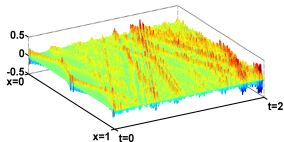
(b) Control for  $Y_2^{h,0} = H_{2,hi}^{h,a} Y_2^{h,1} = 0$



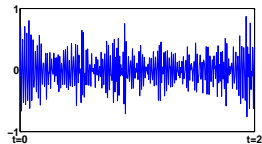
(c) Solution for  $Y_2^{h,0} = H_{2,lo}^{h,o} Y_2^{h,1} = 0$



(d) Control for  $Y_2^{h,0} = H_{2,lo}^{h,o} Y_2^{h,1} = 0$



(e) Solution for  $Y_2^{h,0} = H_{2,hi}^{h,o} Y_2^{h,1} = 0$



(f) Control for  $Y_2^{h,0} = H_{2,hi}^{h,o} Y_2^{h,1} = 0$

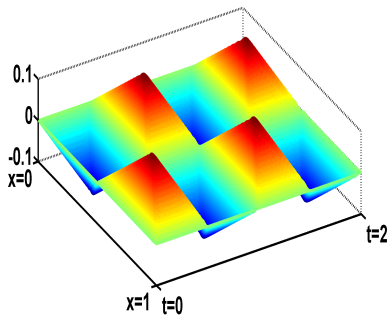
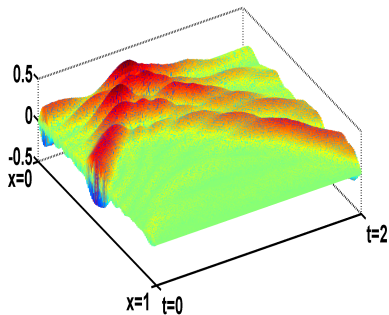


Figure: Typical solution of the adjoint problem (9) corresponding to the minimizer  $(\tilde{\mathbf{U}}_p^{h,0}, \tilde{\mathbf{U}}_p^{h,1})$  of  $\mathcal{J}_p^h$  over  $\mathcal{V}_p^h$  (left) or over  $(\mathcal{B}_p^h \times \mathcal{B}_p^h) \times \mathcal{V}_p^h$  (right).



## Implementation of the conjugate gradient with filtering

Modifications on **Step 2**:  $\mathbf{F}_2^{h,0} = -M_2^h \mathbf{Y}_{2,0,t}^h(T)$  and  $\mathbf{F}_2^{h,1} = M_2^h \mathbf{Y}_{2,0}^h(T)$ .

The Gateaux derivative of  $\mathcal{J}_2^h$  at  $(\mathbf{U}_{2,0}^{h,0}, \mathbf{U}_{2,0}^{h,1})$  is:

$$\mathcal{J}_2^{h,\prime}(\mathbf{U}_{2,0}^{h,0}, \mathbf{U}_{2,0}^{h,1})(\mathbf{U}_2^{h,0}, \mathbf{U}_2^{h,1}) = (\mathbf{F}_2^{h,0}, \mathbf{U}_2^{h,0})_{2,e} + (\mathbf{F}_2^{h,1}, \mathbf{U}_2^{h,1})_{2,e} = (\mathbf{G}_{2,0}^{h,0}, \mathbf{U}_2^{h,0})_{h,1,2} + (\mathbf{G}_{2,0}^{h,1}, \mathbf{U}_2^{h,1})_{h,0,2}.$$

First restriction operator  $\Pi$ :  $(\Pi \mathbf{E}_2^h)_j = E_{2,2j}$ , for all  $1 \leq j \leq (N-1)/2$ .

When both  $(\mathbf{U}_2^{h,0}, \mathbf{U}_2^{h,1})$  and  $(\mathbf{G}_{2,0}^{h,0}, \mathbf{G}_{2,0}^{h,1})$  belong to  $\mathcal{B}_2^h \times \mathcal{B}_2^h$

$$(\mathbf{G}_{2,0}^{h,0}, \mathbf{U}_2^{h,0})_{h,1,2} + (\mathbf{G}_{2,0}^{h,1}, \mathbf{U}_2^{h,1})_{h,0,2} = (\Pi \mathbf{G}_{2,0}^{h,0}, \Pi \mathbf{U}_2^{h,0})_{2h,1,1} + (\Pi \mathbf{G}_{2,0}^{h,1}, \Pi \mathbf{U}_2^{h,1})_{2h,0,1}.$$

The second restriction operator  $\Gamma$

$$(\Gamma \mathbf{E}_2^h)_j = E_{2,2j} + 3(E_{2,2j+1/2} + E_{2,2j-1/2})/4 + (E_{2,2j+1} + E_{2,2j-1})/2 + (E_{2,2j+3/2} + E_{2,2j-3/2})/4.$$

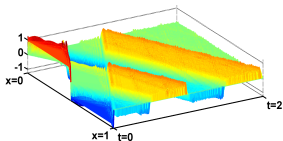
$$(\mathbf{F}_2^{h,0}, \mathbf{U}_2^{h,0})_{2,e} + (\mathbf{F}_2^{h,1}, \mathbf{U}_2^{h,1})_{2,e} = ((S_1^{2h})^{-1} \Gamma \mathbf{F}_2^{h,0}, \Pi \mathbf{U}_2^{h,0})_{2h,1,1} + ((M_1^{2h})^{-1} \Gamma \mathbf{F}_2^{h,1}, \Pi \mathbf{U}_2^{h,1})_{2h,0,1}$$

The two components of the gradient are explicitly given by

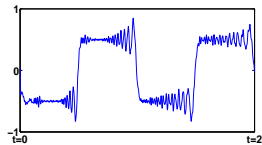
$$\mathbf{G}_{2,0}^{h,0} = \Pi^{-1} (S_1^{2h})^{-1} \Gamma \mathbf{F}_2^{h,0} \quad \text{and} \quad \mathbf{G}_{2,0}^{h,1} = \Pi^{-1} (M_1^{2h})^{-1} \Gamma \mathbf{F}_2^{h,1},$$

where  $\Pi^{-1}$  is the inverse of the restriction operator  $\Pi$  defined as the linear interpolation on a grid of size  $h/2$  of a function defined on a grid of size  $2h$ .

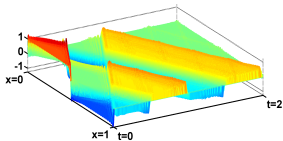
# Numerical results (II)



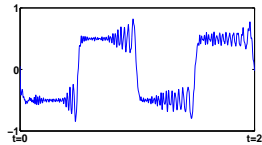
(a) Solution for  $\mathbf{Y}_1^{h,0} = \mathbf{H}_1^h, \mathbf{Y}_1^{h,1} = 0$



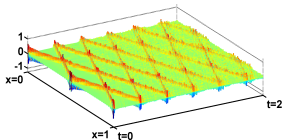
(b) Control for  $\mathbf{Y}_1^{h,0} = \mathbf{H}_1^h, \mathbf{Y}_1^{h,1} = 0$



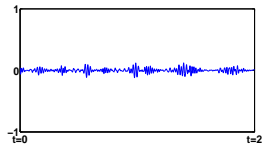
(c) Solution for  $\mathbf{Y}_1^{h,0} = \mathbf{H}_{1,lo}^h, \mathbf{Y}_1^{h,1} = 0$



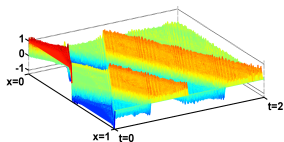
(d) Control for  $\mathbf{Y}_1^{h,0} = \mathbf{H}_{1,lo}^h, \mathbf{Y}_1^{h,1} = 0$



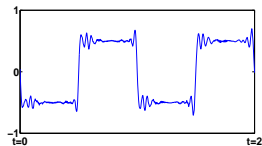
(e) Solution for  $\mathbf{Y}_1^{h,0} = \mathbf{H}_{1,hi}^h, \mathbf{Y}_1^{h,1} = 0$



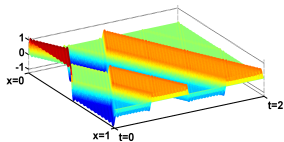
(f) Control for  $\mathbf{Y}_1^{h,0} = \mathbf{H}_{1,hi}^h, \mathbf{Y}_1^{h,1} = 0$



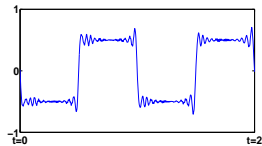
(a) Solution for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_2^h, \mathbf{Y}_2^{h,1} = 0$



(b) Control for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_2^h, \mathbf{Y}_2^{h,1} = 0$

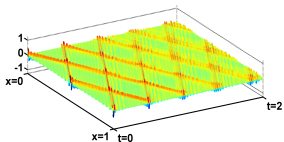


(c) Solution for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_{2,lo}^{h,a}, \mathbf{Y}_2^{h,1} = 0$

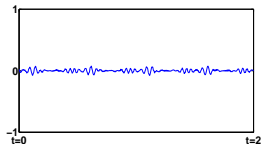


(d) Control for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_{2,lo}^{h,a}, \mathbf{Y}_2^{h,1} = 0$

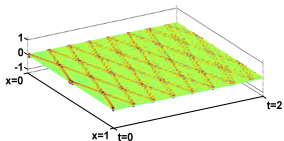
Figure: Solutions of the controlled problem (13) and the corresponding numerical controls for  $p = 2$  arising by minimizing  $\mathcal{J}_2^h$  over the whole space  $\mathcal{V}_2^h$ .



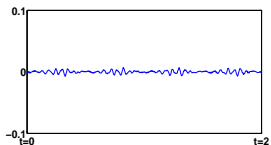
(a) Solution for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_{2,hi}^{h,a}, \mathbf{Y}_2^{h,1} = 0$



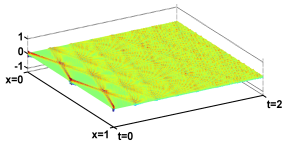
(b) Control for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_{2,hi}^{h,a}, \mathbf{Y}_2^{h,1} = 0$



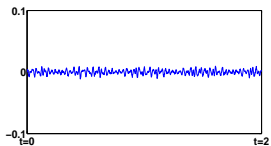
(c) Solution for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_{2,lo}^{h,o}, \mathbf{Y}_2^{h,1} = 0$



(d) Control for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_{2,lo}^{h,o}, \mathbf{Y}_2^{h,1} = 0$



(e) Solution for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_{2,hi}^{h,o}, \mathbf{Y}_2^{h,1} = 0$



(f) Control for  $\mathbf{Y}_2^{h,0} = \mathbf{H}_{2,hi}^{h,o}, \mathbf{Y}_2^{h,1} = 0$

## Conclusions and open problems

In this talk:

- we show numerically the high frequency pathological effects of the  $P_2$  approximation of the boundary controllability problem for the  $1 - d$  wave equation we discovered from a theoretical point of view in [4].
- we also illustrate the efficiency of our bi-grid filtering algorithm in recovering the convergence of the numerical controls and compare our numerical results with the ones for the  $P_1$  approximation.
- our conclusion is that after restricting the space over which we minimize the discrete functionals to the bi-grid one, we obtain more accurate controls for the quadratic approximation than for the linear one.
- the same analysis can be done for the DG method in [3].
- the filtering technique can be generalized to higher order finite elements approximation of waves ( $p \geq 3$ ) on uniform meshes, a higher and higher accuracy of the numerical controls being expected.

Open problems:

- the high frequency effects of the numerical approximations on irregular meshes is a completely unknown open problem.
- higher-order FEM and DG approximations for other models like the Schrödinger equation.

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Thank you very much for your attention!