

A collocated MOR for advective-dominant PDEs: hyper-reduction of ADER scheme on evolving chimera meshes

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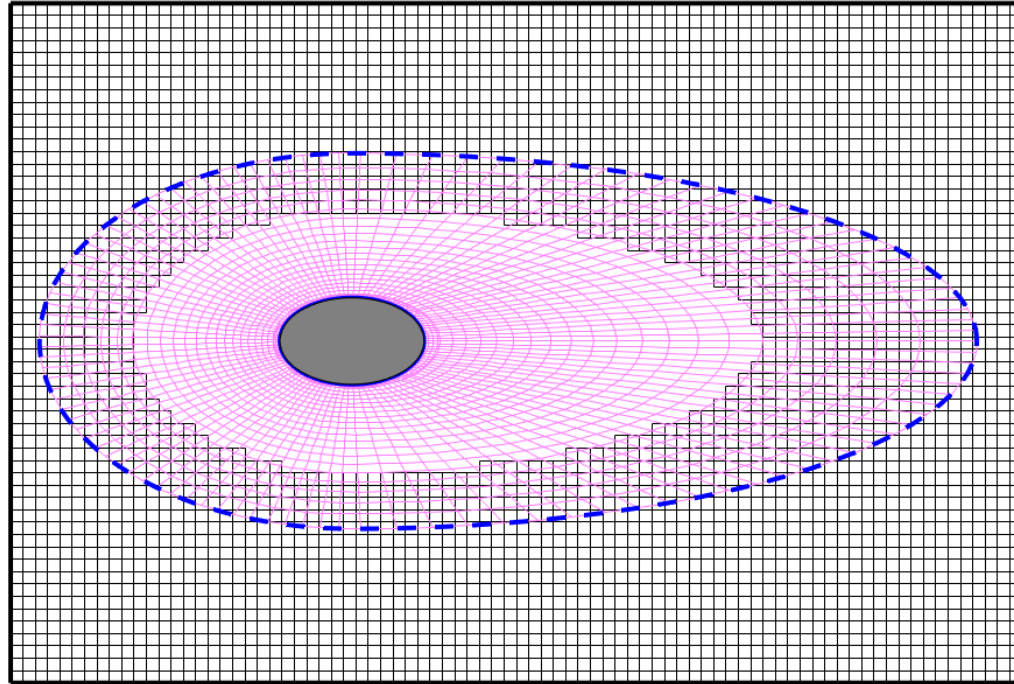
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Outline

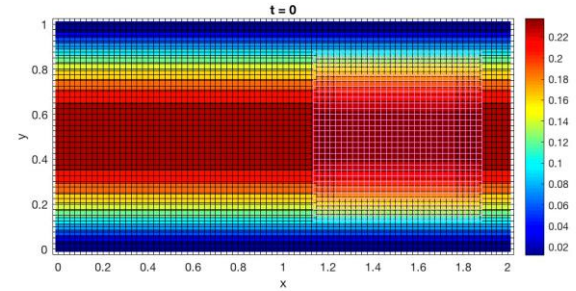
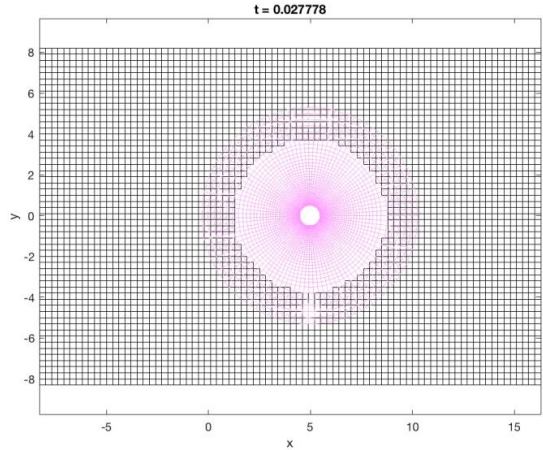
- Chimera mesh and cell classification
- The high-fidelity method
- The hyper-reduced approach
- Beyond the Kolmogorov n -width
- Numerical results
- Conclusions and Future Perspectives



The Overset Grid or Chimera Mesh



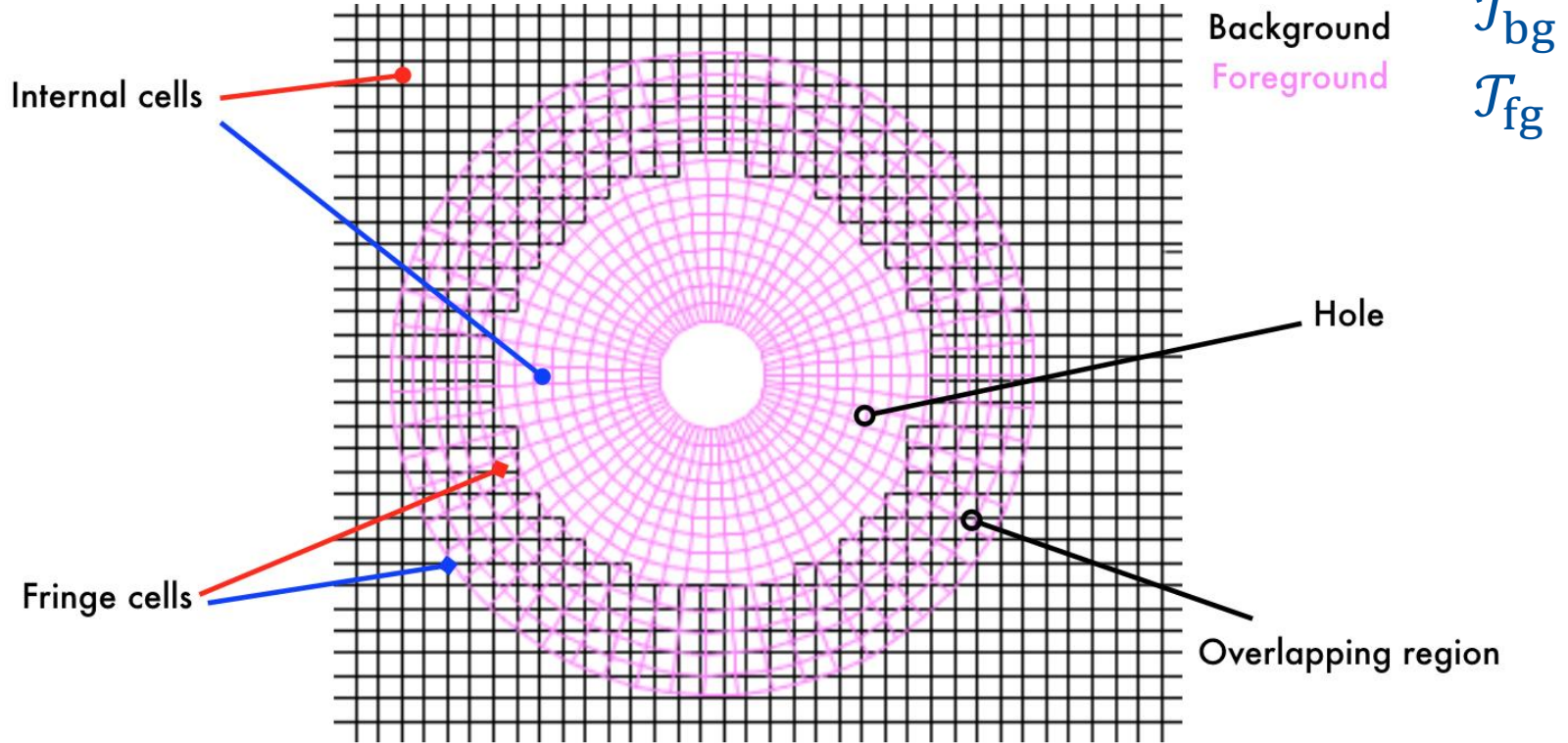
The Overset Grid or Chimera Mesh



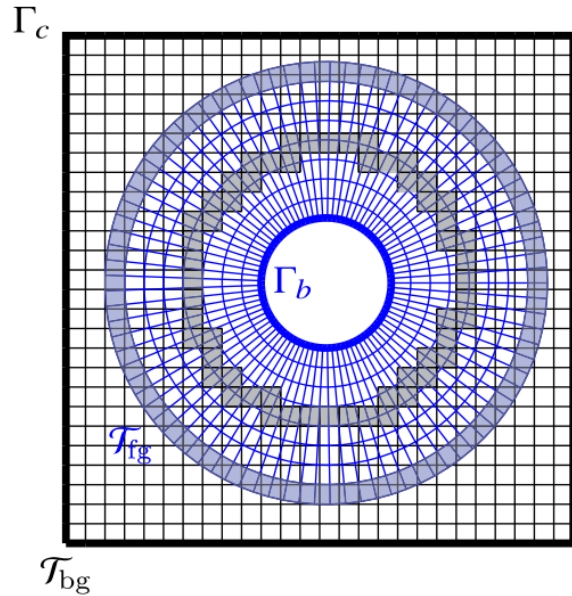
Pro: high fidelity solution wrt the real geometry of the physical phenomenon

Cons: i) Communication among different meshes;
ii) Mesh movement and deformation

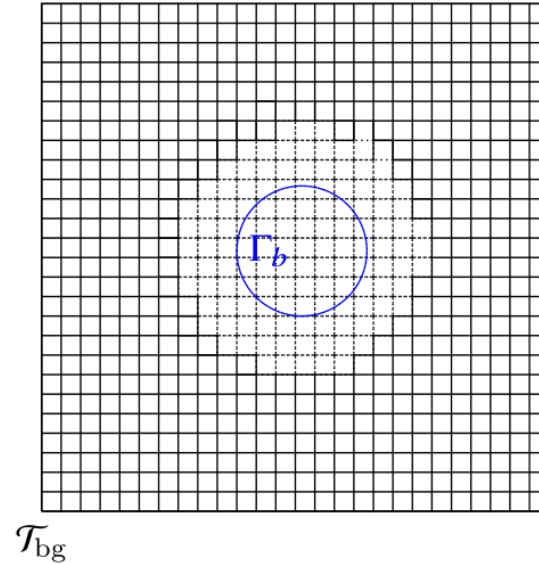
A Cell Classification



Another cell classification

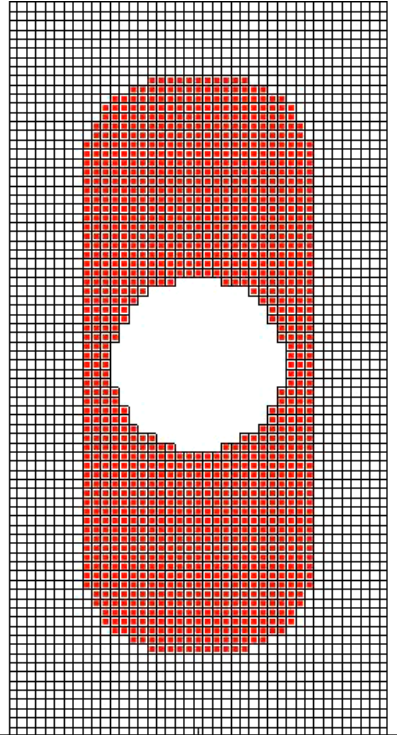


$\mathcal{T}_{bg}^{hole}(t)$



$$\mathcal{T}_{bg}(t) = \mathcal{T}_{bg}^{hole}(t) \cup \mathcal{T}_{bg}^{ol}(t) \cup \mathcal{T}_{bg}^{nol}(t)$$

Another cell classification



$$\bigcup_t \mathcal{T}_{bg}(t) = \bigcup_t \mathcal{T}_{bg}^{ol}(t) \cup \bigcup_t \mathcal{T}_{bg}^{nol}(t) = \mathcal{T}_{bg}^{HF} \cup \mathcal{T}_{bg}^{ROM}$$

$$\mathcal{T}_{bg}^{hole}(t) \subset \mathcal{T}_{bg}^{HF} \quad \forall t$$

Static partitions: \mathcal{T}_{bg}^{ROM} and \mathcal{T}_{fg} .

Dynamic partition: \mathcal{T}_{bg}^{HF} .



The high-fidelity method

Find $\mathbf{u}: \Omega(t) \times (0, T] \rightarrow \mathbb{R}^\delta$ such that

$$\mathbf{F}(\mathbf{u}, \nabla \mathbf{u}) = \boldsymbol{\beta}(\mathbf{u}) - \nu(\mathbf{u}) \nabla \mathbf{u}$$

$$\partial_t \mathbf{u} + \nabla \cdot \mathbf{F}(\mathbf{u}, \nabla \mathbf{u}) = \mathbf{f}$$

Properly closed by boundary conditions and initial conditions.

Motion equation for $\mathbf{X}_{fg} \in \Omega_{fg}$:

$$\begin{cases} \dot{\mathbf{X}} = \mathbf{V}(\mathbf{x}, t; \mathbf{u}) \\ \mathbf{X}(0) = \mathbf{X}_0 \end{cases}$$

Arbitrary high-order DERivatives: ADER (FV version)

Prediction

- From solution at $t = t^n$: (spatial) local \mathcal{P}_2^{loc} -interpolation over any cell $\Omega_i^n = \Omega_i(t^n)$.
- Local-space time solution over any $\mathcal{C}_i^n = \Omega_i(t) \times (t^n, t^{n+1})$.

Correction

- FV scheme by integrating the PDE over \mathcal{C}_i^n .
- Flux discretization by employing the local space-time predictor

Overset mesh

- Cell management of the overlapping zone

¹Dumbser *et al.*, SIAM J. Sci. Comp. (2017); Titaterev & Toro, J. Comp. Phys. (2005); Bergmann *et al.*, SIAM J. Sci. Comp. (2021); Bergmann *et al.*, J. Sci. Comp. (2022).

ADER: Prediction

Space-time cell: $\mathcal{C}_i^n = \Omega_i(t) \times (t^n, t^{n+1})$.

From the mesh motion equation:

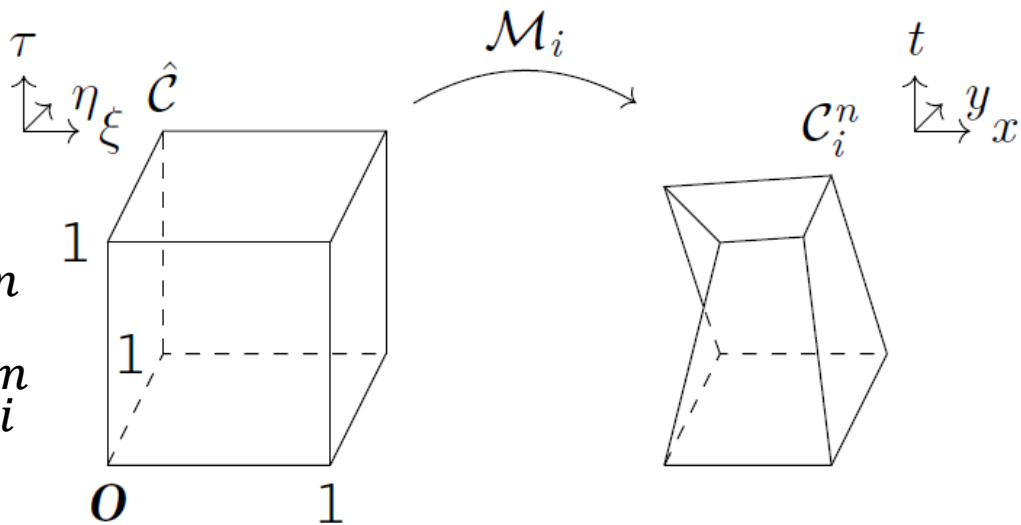
$$\mathcal{M}_i: \begin{cases} x = x(\xi, \eta, \tau) \\ y = y(\xi, \eta, \tau) \\ t = t^n + \Delta t \tau \end{cases}$$

Find the local space-time solution

$$\mathbf{q}: \mathcal{C}_i^n \rightarrow \mathbb{R}^\delta \text{ s. t.}$$

$$\begin{cases} \partial_t \mathbf{q} + \nabla \cdot \mathbf{F}(\mathbf{q}, \nabla \mathbf{q}) = \mathbf{f}; & \text{in } \mathcal{C}_i^n \\ \mathbf{q}|_{t=t^n} = \mathcal{P}_2^{loc} \mathbf{u}^n; & \text{on } \Omega_i^n \end{cases}$$

The equation is solved via FE.



ADER: Correction

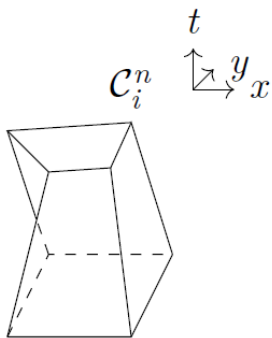
Definition: $\nabla_{x,t} := [\partial_t, \nabla]^T$; $\mathbf{u} := [\mathbf{u}, F(\mathbf{u}, \nabla \mathbf{u})]^T$

$$\partial_t \mathbf{u} + \nabla \cdot F(\mathbf{u}, \nabla \mathbf{u}) = \mathbf{f} \iff \nabla_{x,t} \cdot \mathbf{u} = \mathbf{f}$$

Integration over \mathcal{C}_i^n :

$$-|\Omega_i^n| \mathbf{u}_i^n + |\Omega_i^{n+1}| \mathbf{u}_i^{n+1} + \sum_{j=1}^4 \int_{\Gamma_{ij}^n} \mathbf{u} \cdot \mathbf{n} d\Gamma = \int_{\mathcal{C}_i^n} \mathbf{f} d\mathcal{C}$$

with $\mathbf{u}_i^n = \frac{1}{|\Omega_i^n|} \int_{\Omega_i^n} \mathbf{u}(\mathbf{x}, t^n) d\Omega$.



Rusanov approach¹:

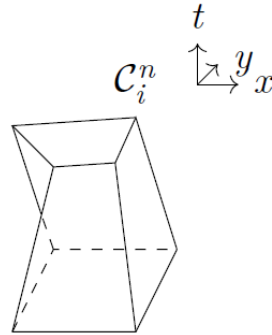
$$[\mathbf{u} \cdot \mathbf{n}]_{\Gamma_{ij}^n} \approx \frac{1}{2} (\mathbf{u}_j^+ + \mathbf{u}_j^-) \cdot \mathbf{n}_{x,t} - \frac{S}{2} (\mathbf{q}_j^+ - \mathbf{q}_j^-)$$

¹Bergmann *et al.*, SIAM J. Sci. Comp. (2021).

Stabilization of the scheme

Let $\tilde{n} = \mathbf{n} / \sqrt{n_x^2 + n_y^2}$.

$$S_{AD} = \max |\rho(\mathcal{A})| = \frac{1}{2} \left| \tilde{\sigma} + \sqrt{\left(a_x^2 + \frac{4\nu}{\varepsilon}\right) \tilde{n}_x^2 + 2a_x a_y \tilde{n}_x \tilde{n}_y + \left(a_y^2 + \frac{4\nu}{\varepsilon}\right) \tilde{n}_y^2} \right|$$



Optimal relaxation parameter ε

Corollary: Optimal relaxation time¹

For a given mesh size h and a numerical scheme of order p for $u_{hyp,h}$ solving the hyperbolized problem derived by the original parabolic problem, the optimal relaxation time ε_p is

$$\varepsilon_p = \mathcal{O}(1) \frac{h^p}{C_p}$$

Chosen relaxation time: $\varepsilon = \varepsilon_2/2$.

¹Toro & Montecinos, SIAM J. Sci. Comp. (2014); Carlino, PhD Thesis (2021); Bergmann *et al.*, SIAM J. Sci. Comp. (2021)

The advective-dominant limit

For a hyperbolic equation ($\nu = 0$), the stabilization coefficient is the maximum eigenvalue of the ALE Jacobian matrix

$$A_{\tilde{\mathbf{n}}}^V = \sqrt{n_x^2 + n_y^2} \left(\frac{\partial F}{\partial u} \tilde{\mathbf{n}}_x - \mathbf{V} \cdot \tilde{\mathbf{n}} \mathbf{I} \right)$$

Namely:

$$s_A = \max |\rho(A_{\tilde{\mathbf{n}}}^V)| = |\mathbf{a} \cdot \tilde{\mathbf{n}}_x + \tilde{n}_t|$$

Proposition: Limit of vanishing diffusion¹

$$\lim_{\nu \rightarrow 0} s_{AD} = \frac{1}{2} |\tilde{\sigma} + \mathbf{a} \cdot \tilde{\mathbf{n}}_x + n_t| = s_A$$

¹Carlino, PhD Thesis (2021); Bergmann *et al.*, SIAM J. Sci. Comp. (2021)

The hyper-reduced approach

Find $\mathbf{u}: \Omega(t) \times (0, T] \times \mathcal{P} \rightarrow \mathbb{R}^\delta$, with $\mathcal{P} \subset \mathbb{R}^p$, such that

$$\partial_t \mathbf{u}(\mathbf{z}) + \nabla \cdot \mathbf{F}(\mathbf{u}(\mathbf{z}), \nabla \mathbf{u}(\mathbf{z})) = \mathbf{f}$$

With $\mathbf{z} \in \mathcal{P}$ parameter vector.

Motion equation for $\mathbf{X}_{fg} \in \Omega_{fg}$:

$$\begin{cases} \dot{\mathbf{X}} = \mathbf{V}(\mathbf{x}, t; \mathbf{u}; \mathbf{z}) \\ \mathbf{X}(0) = \mathbf{X}_0 \end{cases}$$



The reduced basis $\text{span}\{\phi_i\}_{i=1}^{N_{\text{ROM}}}$

$$\begin{aligned}\Phi &\in \mathbb{R}^{N \times N_{\text{ROM}}} \\ W &\in \mathbb{R}^{N \times N} \\ \hat{\mathbf{u}}_{\text{HF}} &\in \mathbb{R}^N\end{aligned}$$

It holds

$$u_{\text{ROM}}(\mathbf{x}, t; \mathbf{z}) = \sum_{i=1}^N a_i(t; \mathbf{z}) \phi_i(\mathbf{x}) \Rightarrow \hat{\mathbf{u}}_{\text{ROM}} = \Phi \mathbf{a}$$

and

$$a_i(t; \mathbf{z}) = \int_{\Omega(t; \mathbf{z})} u_{\text{HF}}(\mathbf{x}, t; \mathbf{z}) \phi_i(\mathbf{x}) \, d\mathbf{x} \Rightarrow \mathbf{a} = \Phi^T W \hat{\mathbf{u}}_{\text{HF}}.$$

Thus:

$$\hat{\mathbf{u}}_{\text{ROM}} = \Phi \Phi^T W \hat{\mathbf{u}}_{\text{HF}}$$

W is the matrix collecting the Gauss quadrature weights.

The collocated reduced approach

The quadrature weights matrix W can be replaced by an empirical quadrature weights matrix \tilde{W} . This new matrix is found by NNLS wrt the integral

$$\int_{\Omega(t; \mathbf{z})} u_{\text{HR}}(\mathbf{x}, t; \mathbf{z}) \phi_i(\mathbf{x}) \, d\mathbf{x}$$

for any $i = 1, \dots, N$, $\mathbf{z} \in Z$ and $(\mathbf{x}, t) \in \Omega(t; \mathbf{z}) \times [0, T]$.

Thus the projection reads

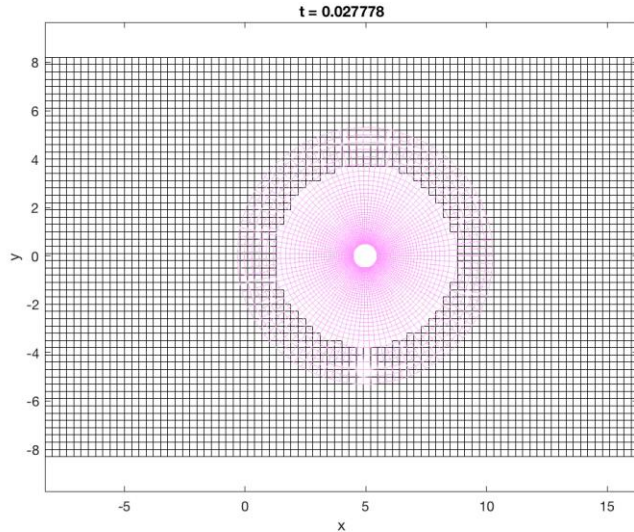
$$\hat{\mathbf{u}}_{\text{HR}} = \Phi \Phi^T \tilde{W} \hat{\mathbf{u}}_{\text{HF}}$$

$$\Phi \in \mathbb{R}^{N_{\text{HR}} \times N_{\text{ROM}}}$$

$$\tilde{W} \in \mathbb{R}^{N_{\text{HR}} \times N_{\text{HR}}}$$

$$\hat{\mathbf{u}}_{\text{HF}} \in \mathbb{R}^{N_{\text{HR}}}$$

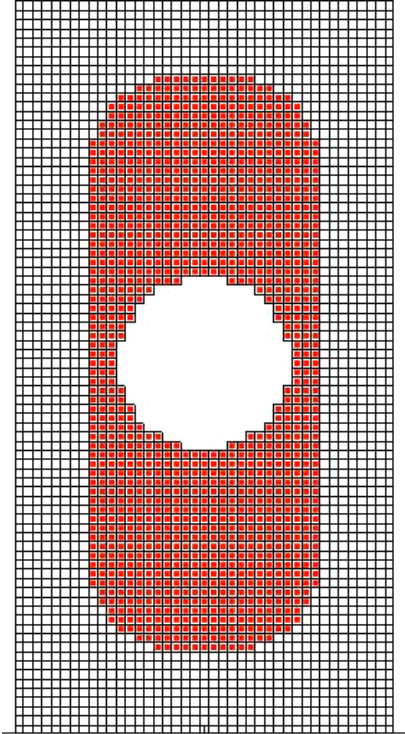
The issue of the background mesh



A domain decomposition approach.
Two reduced basis:

$$\text{span} \left\{ \phi_i^{\text{bg}} \right\}_{i=1}^{N_{\text{ROM}}^{\text{bg}}} \text{ on } \mathcal{T}_{\text{bg}}^{\text{ROM}} = \cup_{t,z} \mathcal{T}_{\text{bg}}^{\text{mol}}$$
$$\text{span} \left\{ \phi_i^{\text{fg}} \right\}_{i=1}^{N_{\text{ROM}}^{\text{fg}}} \text{ on } \mathcal{T}_{\text{fg}}$$

The issue of the background mesh



On $\mathcal{T}_{bg}^{HF} = U_{t,z} \mathcal{T}_{bg}^{ol}$ the recovered solution is HF.

Beyond the Kolmogorov n -width

Let $\varphi: \mathcal{P} \rightarrow \Omega$ be the mapping associating any $\mathbf{z} \in \mathcal{P}$ to a parametric solution $\mathbf{u} = \varphi(\mathbf{z}) \in \Omega$ under the constraint given by the PDE:

$$\mathcal{R}(u(\mathbf{z}), \mathbf{z}) = 0 \quad \forall \mathbf{z} \in \mathcal{P}$$

The residual implicitly define a manifold in the Hilbert space $(V, \|\cdot\|)$. Thus $\text{im}\varphi \subseteq V$.

Let V_n be a n -dimensional ($n < +\infty$) subspace of V . It holds

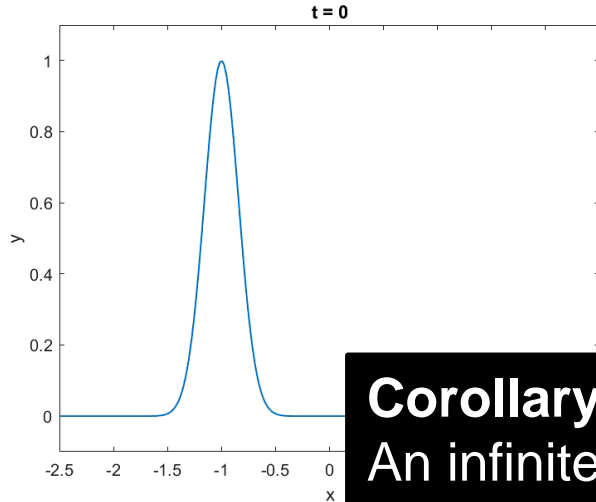
$$d_{V_n} = \inf_{v \in \text{im}\varphi} \sup_{v_n \in V_n} \|v - v_n\| \geq d_n(\text{im}\varphi) = \sup_{\substack{W_n \subseteq V \\ n = \dim W_n < +\infty}} \inf_{w \in \text{im}\varphi} \sup_{w_n \in W_n} \|w - w_n\|$$

With W_n a generic n -dimensional subspace of V .



An example: the advection of a compact support function

Fourier \cong SVD at fixed time

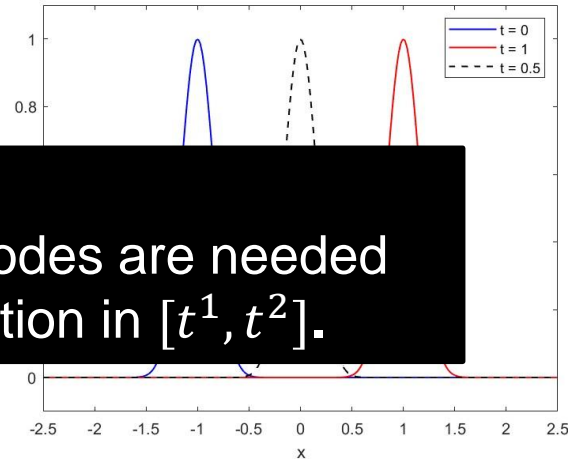


Proposition

There not exists a linear combination representing the evolution of the solution from t^1 to t^2 .

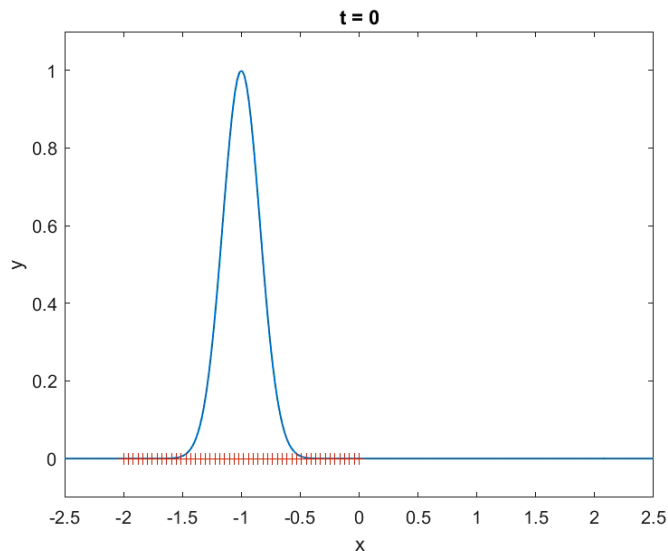
Corollary

An infinite number of modes are needed for represeing the evolution in $[t^1, t^2]$.



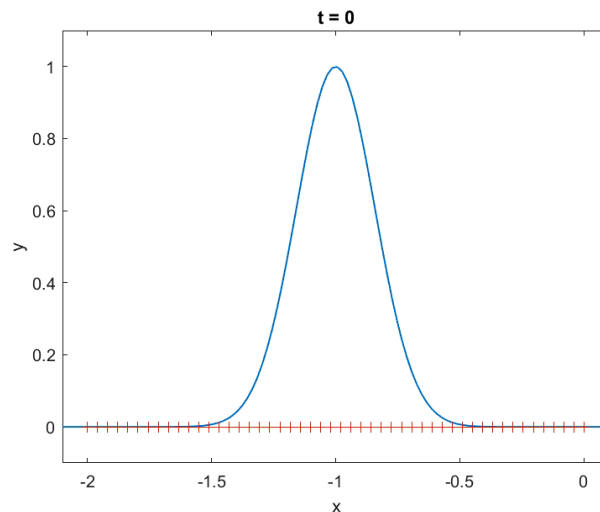
$$f(t = 0.5) \neq a_0 f(t = 0) + a_1 f(t = 1)$$

Reference Mapping vs Relativity of the frame



φ_{Ω}

POV: the foreground mesh



$\varphi_{\mathcal{T}_{bg}^{HR}}; \varphi_{\mathcal{T}_{bg}^{ROM}}; \varphi_{\mathcal{T}_{fg}}$

$$d_n(\text{im}\varphi_{\Omega}) \geq d_n(\text{im}\varphi_{\mathcal{T}_{bg}^{HR}}) + d_n(\text{im}\varphi_{\mathcal{T}_{bg}^{ROM}}) + d_n(\text{im}\varphi_{\mathcal{T}_{fg}})$$



Numerical results

Performance analysis: Hyper-Reduced (HR) vs High-Fidelity (HF).

$$\text{Err} = \frac{\sqrt{\int_0^T \int_{\Omega} (u_{\text{HR}} - u_{\text{HF}})^2 dx dt}}{\sqrt{\int_0^T \int_{\Omega} u_{\text{HF}}^2 dx dt}}$$

$$u_{\text{HR}} \propto \varepsilon_{\text{NNLS}}$$

$$\varepsilon_{\text{NNLS}} = 0 \Rightarrow u_{\text{HR}} = u_{\text{ROM}}$$

Hyper-reduction analysis:

Total number of cells: N

Number of cells in \mathcal{T}_{fg} : N_{fg}

Number of cells in $\mathcal{T}_{\text{bg}}^{\text{HF}}$: $N_{\text{bg}}^{\text{HF}}$

Number of cells in $\mathcal{T}_{\text{bg}}^{\text{ROM}}$: $N_{\text{bg}}^{\text{ROM}}$

Number of cells in \mathcal{T}_{bg} : $N_{\text{bg}} = N_{\text{bg}}^{\text{HF}} + N_{\text{bg}}^{\text{ROM}}$

Number of HR cells in $\mathcal{T}_{\text{bg}}^{\text{ROM}}$: $N_{\text{bg}}^{\text{HR}}$

Number of HR cells in \mathcal{T}_{fg} : $N_{\text{fg}}^{\text{HR}}$

The best approximation: full projection on ROM basis

Let $V_n = \text{span} \left\{ \phi_i^{\text{bg}} \right\}_{i=1}^n \cup \text{span} \left\{ \phi_i^{\text{fg}} \right\}_{i=1}^n \cup \tilde{V}_n \Big|_{\mathcal{T}_{\text{bg}}^{\text{HF}}}$ be the employed ROM subspace.

$$\text{Err} \Big|_{\varepsilon_{\text{NNLS}} \neq 0} \geq \text{Err} \Big|_{\varepsilon_{\text{NNLS}} = 0} \Rightarrow d_{V_n} \Big|_{\varepsilon_{\text{NNLS}} \neq 0} \geq d_{V_n} \Big|_{\varepsilon_{\text{NNLS}} = 0}$$

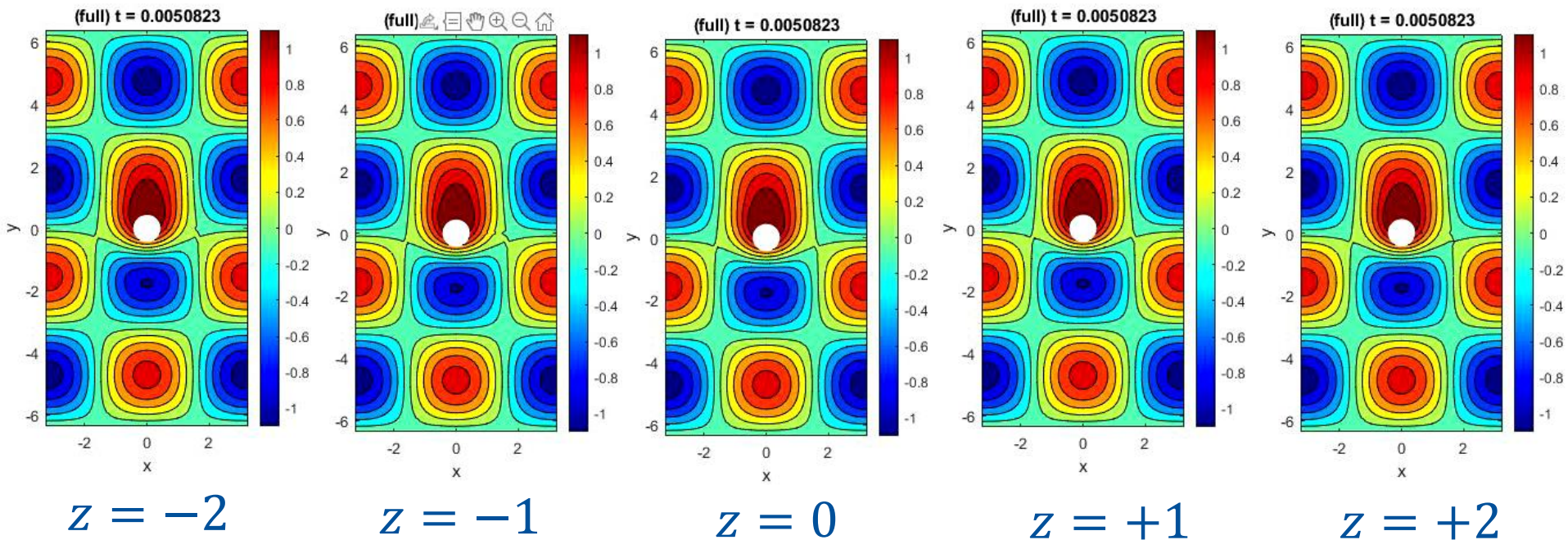
Since ϕ_i^* ($i = 1, \dots, n$, $*$ = bg, fg) is defined all over \mathcal{T}_* , we expect

$$\text{Err} \Big|_{\varepsilon_{\text{NNLS}}^1} \leq \text{Err} \Big|_{\varepsilon_{\text{NNLS}}^2}$$

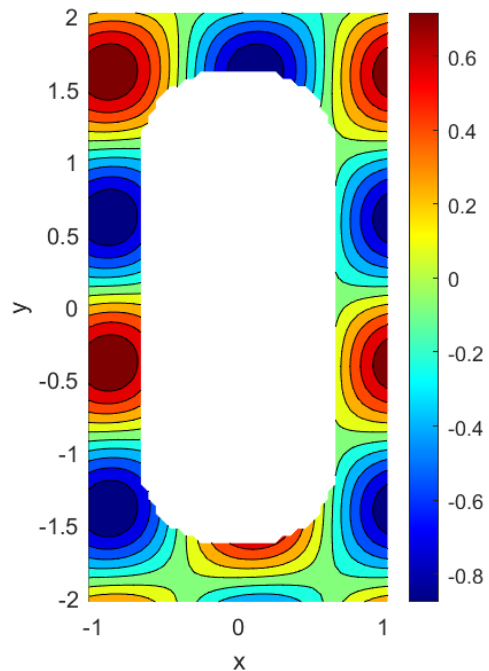
For $\varepsilon_{\text{NNLS}}^1 \leq \varepsilon_{\text{NNLS}}^2$.

Linear test case: $\nu = 0.05$, $\beta = [1,1]^T$, $z \in [-2,2]$

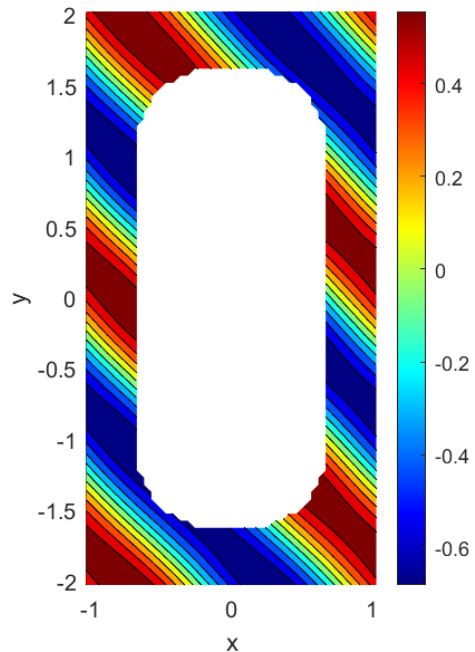
$$u(x, t; z) = \exp[-x^2 - (y + zt)^2 + 0.5^2] + \cos(x - 0.5t) \sin(y - 0.5t)$$



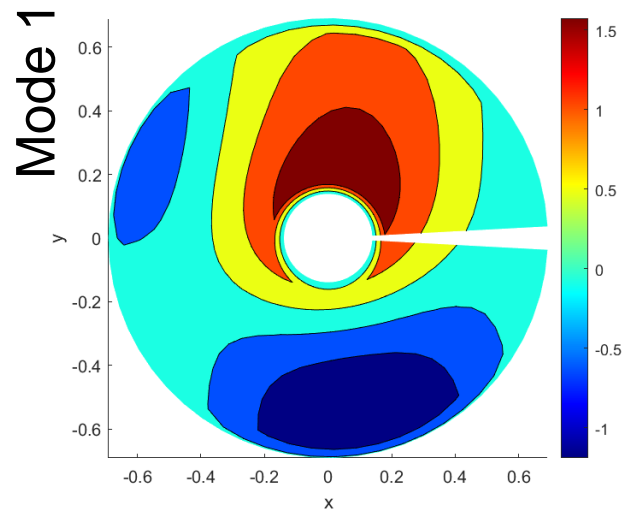
Linear test case



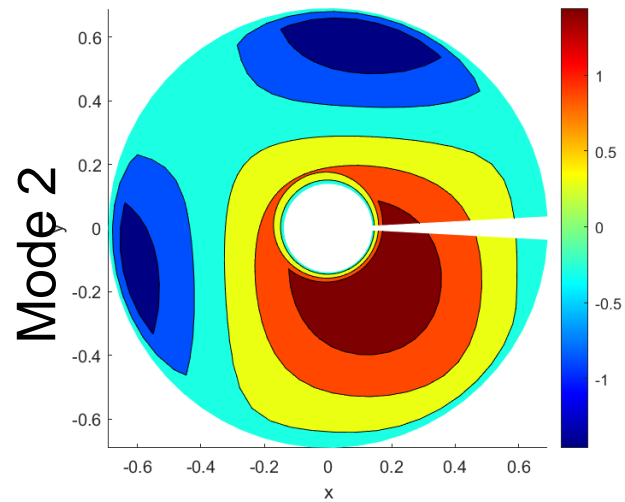
Mode 1



Mode 2



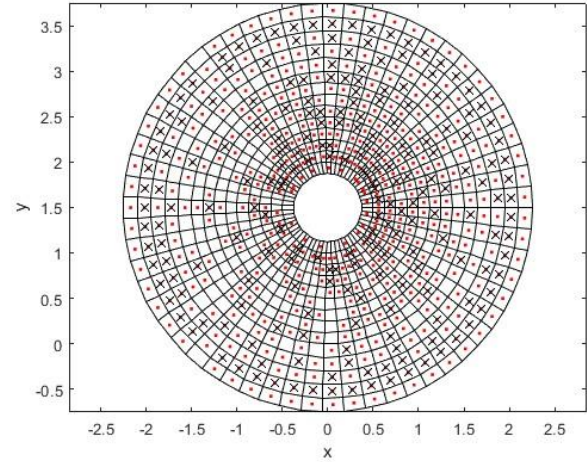
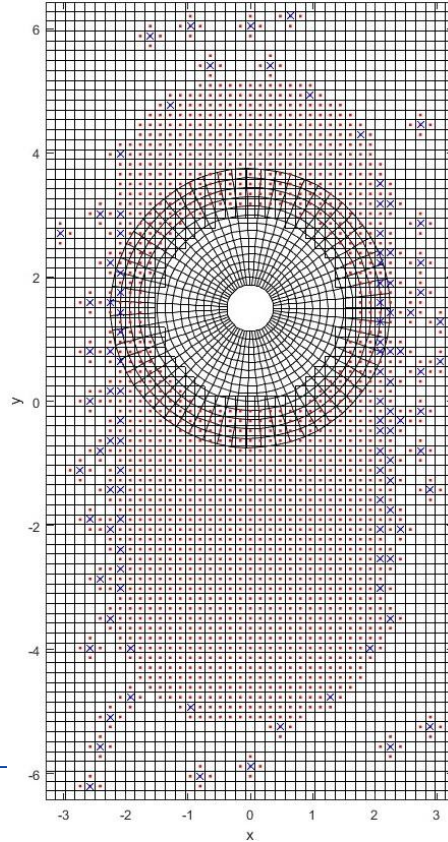
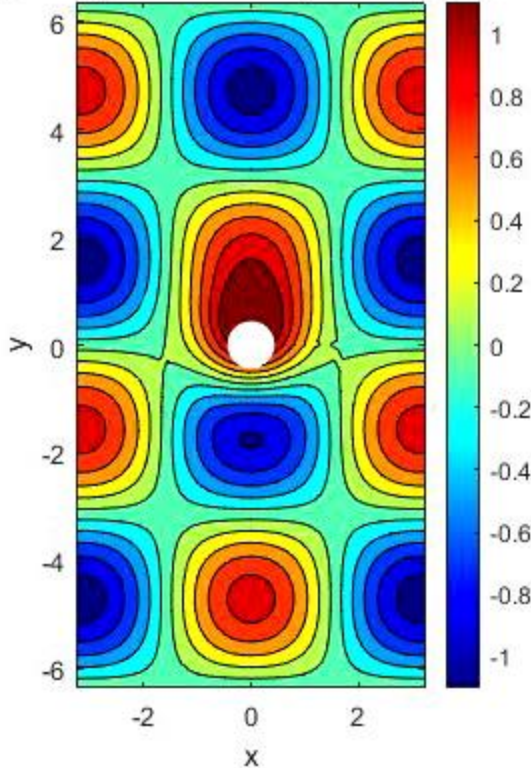
Mode 1



Mode 2

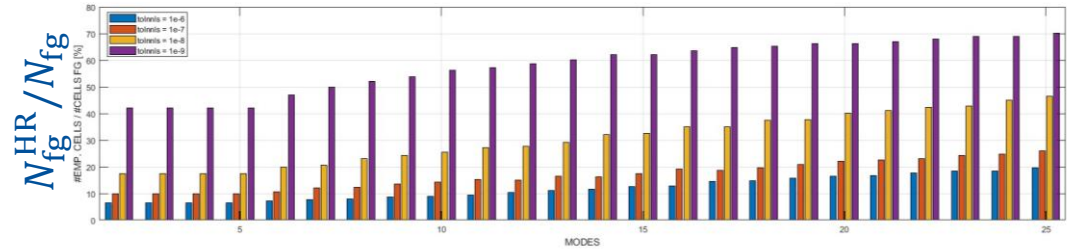
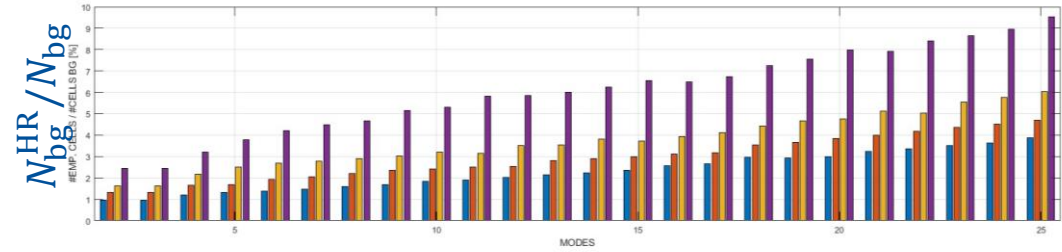
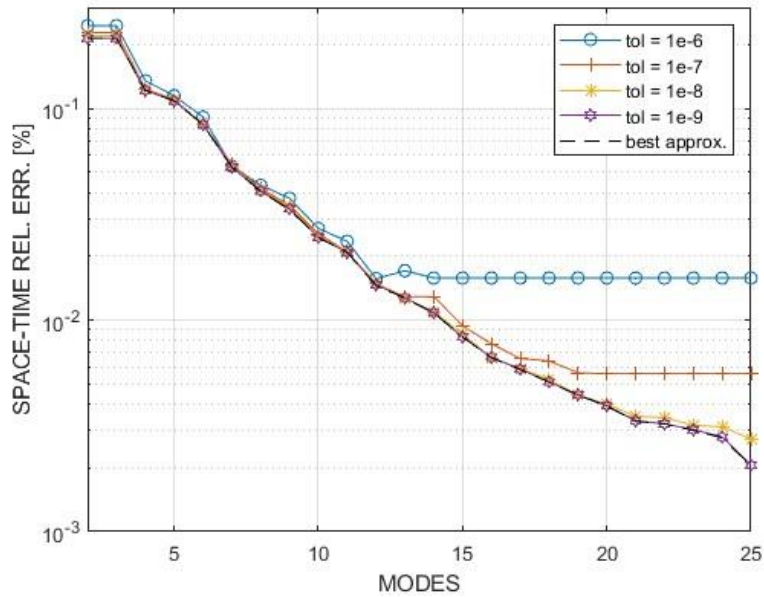
In-sample simulation ($z = +1$)

($N = 10$, $\text{tolnnls} = 1\text{e-}8$) $t = 0.0050823$



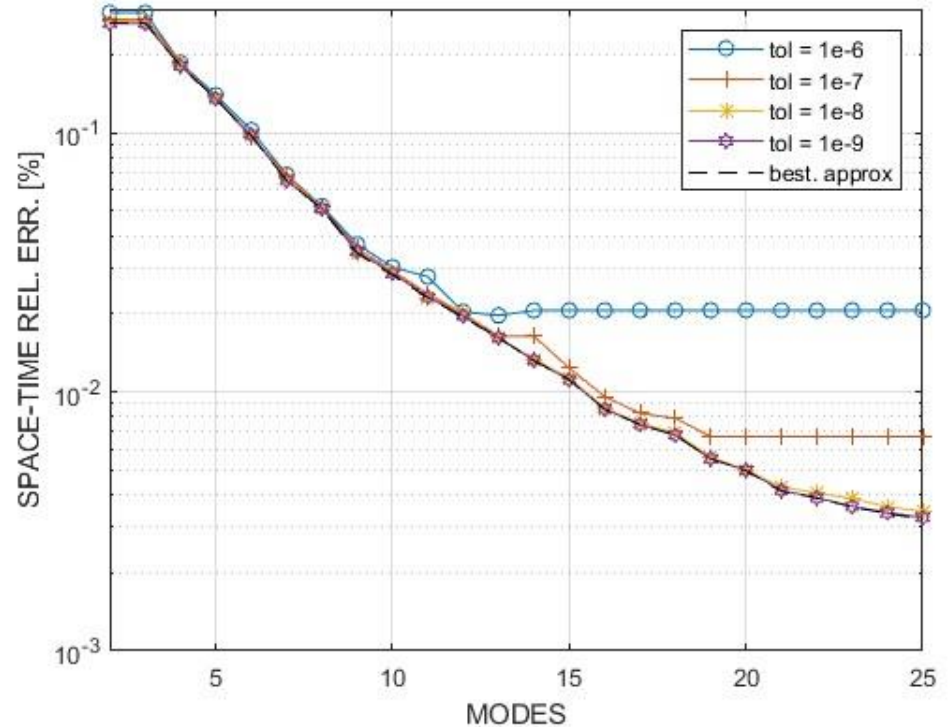
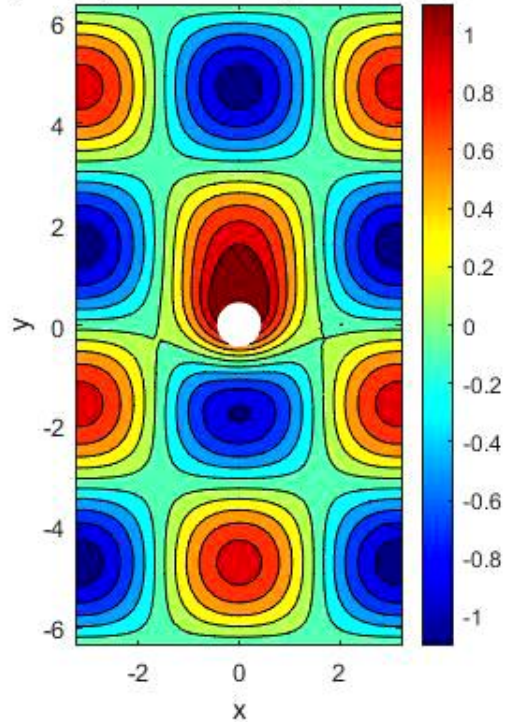
$$N = 4141; N_{bg} = 3321; N_{fg} = 826;$$
$$N_{fg}^{HF} = 1350 \text{ (40\%)}$$

In-sample simulation

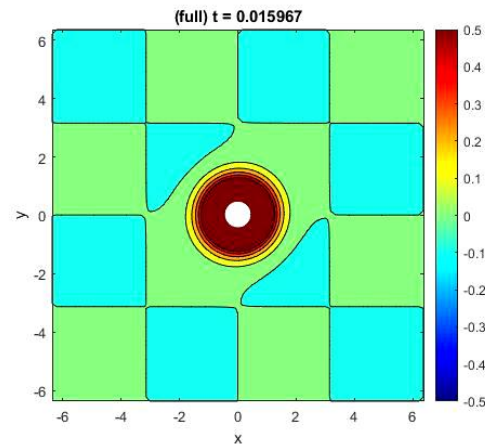
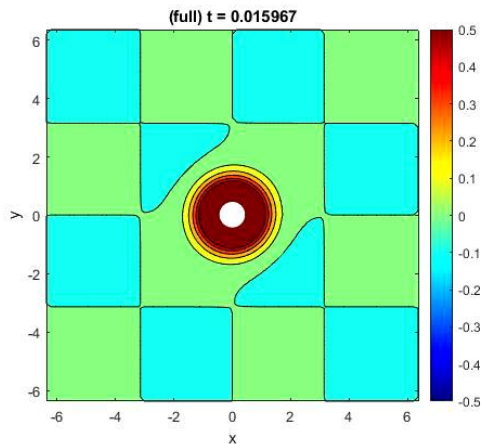
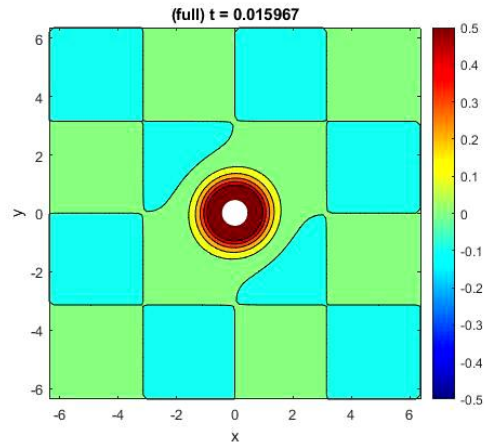
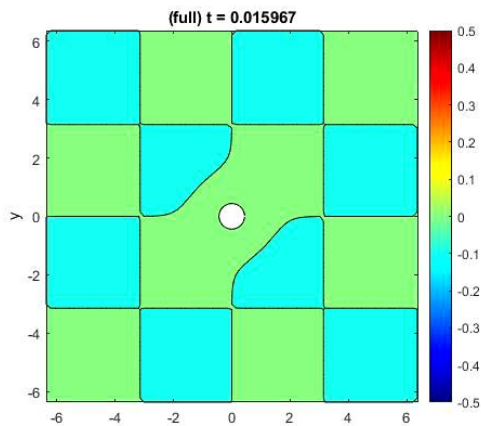


Out-of-sample simulation ($z = +1.7$)

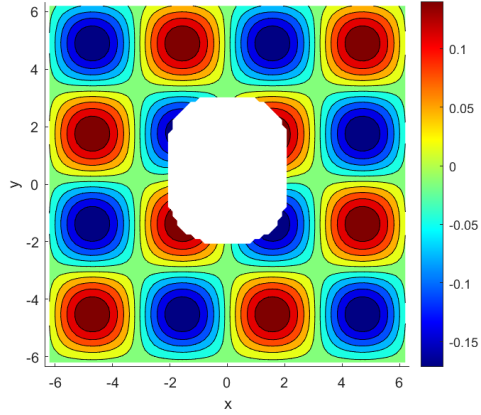
($N = 10$, $\text{tolnnls} = 1\text{e-}8$) $t = 0.0050823$



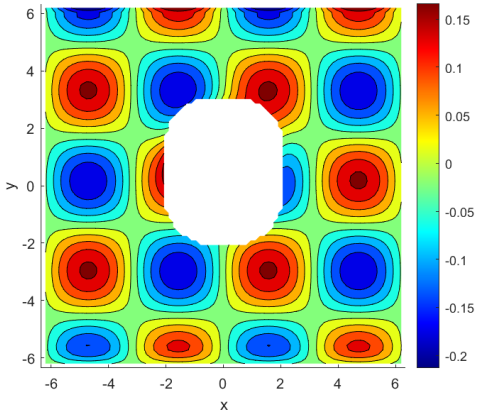
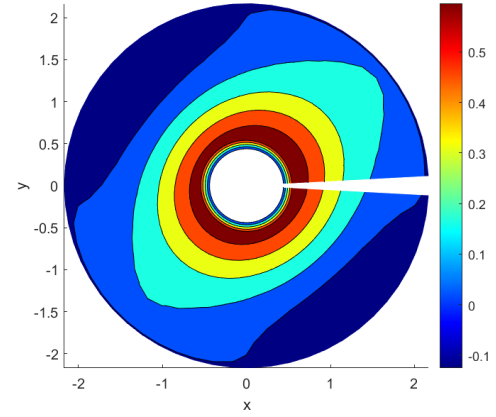
Nonlinear case: $\nu = 0.05$, $\beta = \tanh(u) [0,1]^T$, $z \in [0.1,2]$,
 $f = \sin(x) \sin(y)$



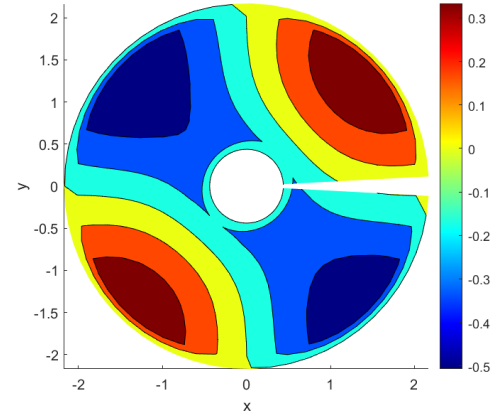
Nonlinear test case



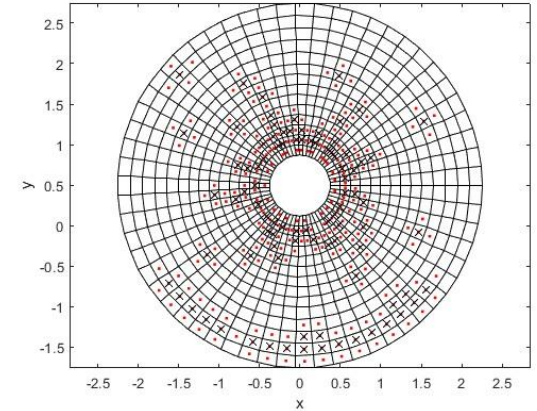
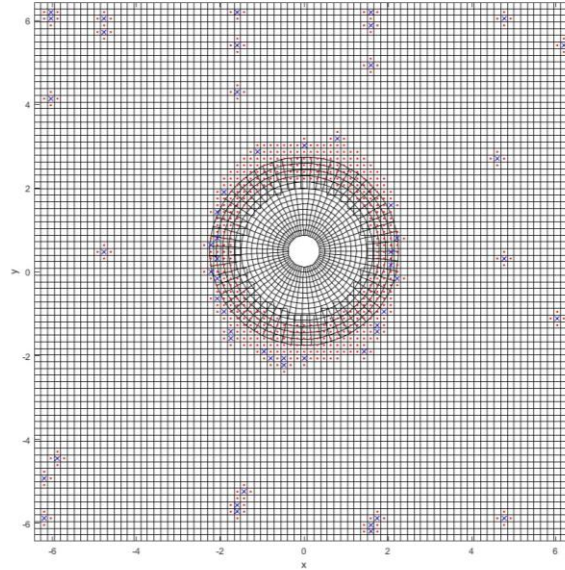
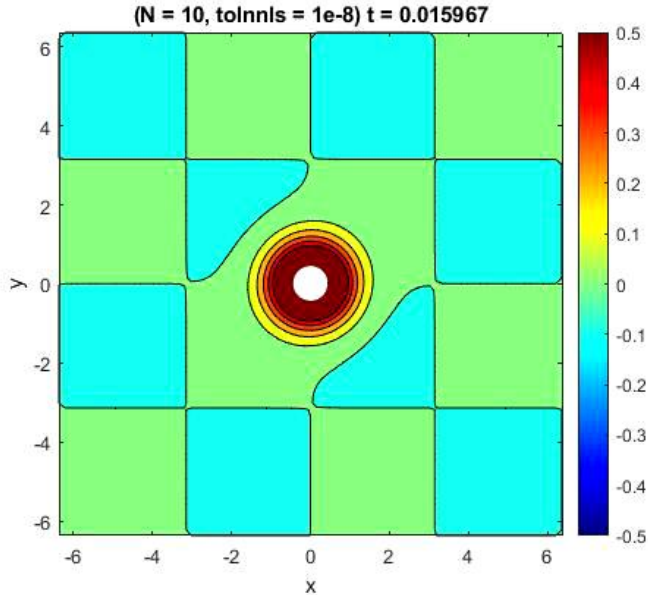
Mode 1



Mode 2

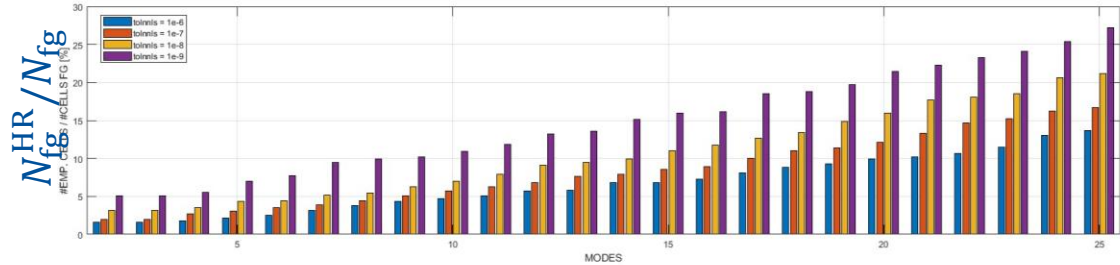
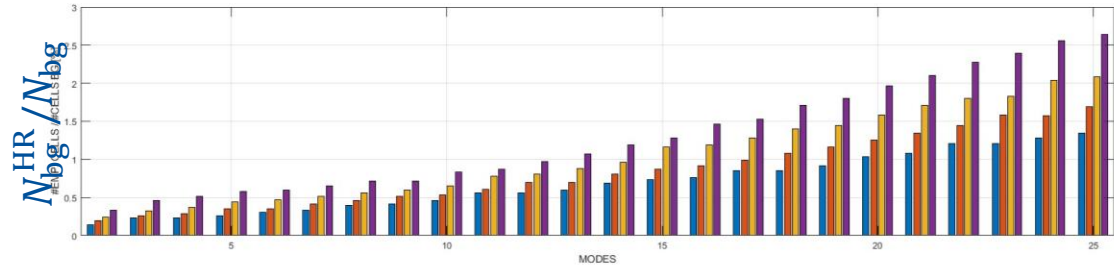
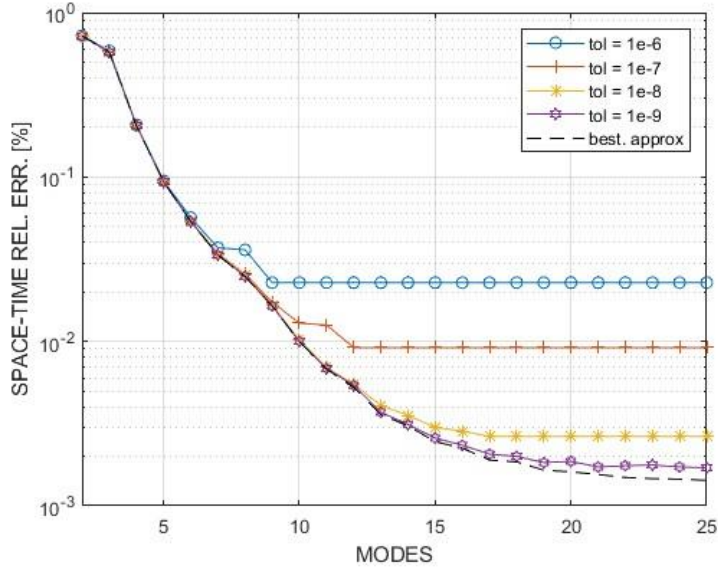


In-sample simulation ($z = +1$)

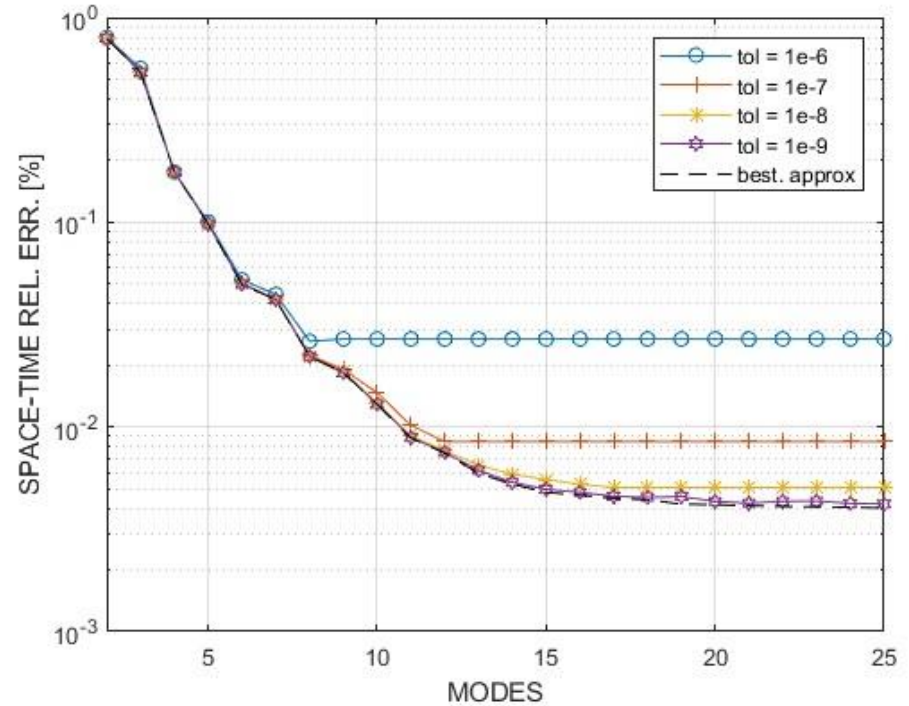
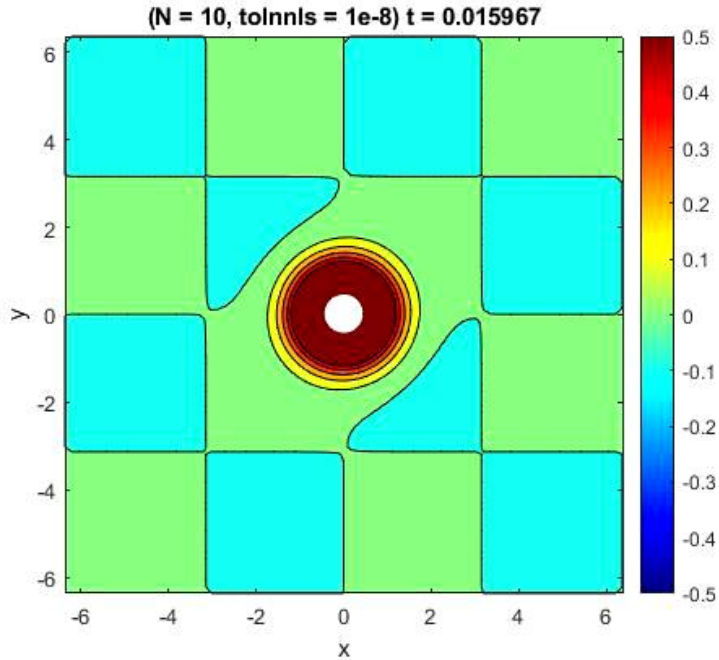


$$N = 7387; N_{bg} = 6561; N_{fg} = 826;$$
$$N_{fg}^{HF} = 491 \text{ (7.5\%)}$$

In-sample simulation

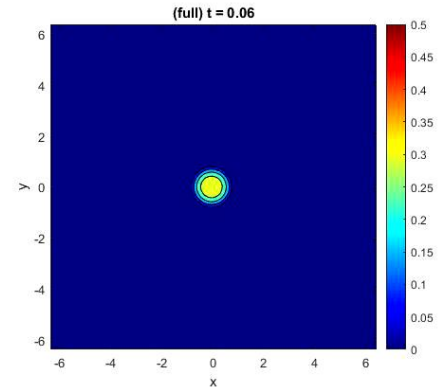
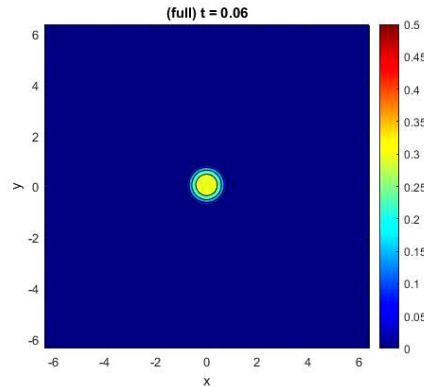
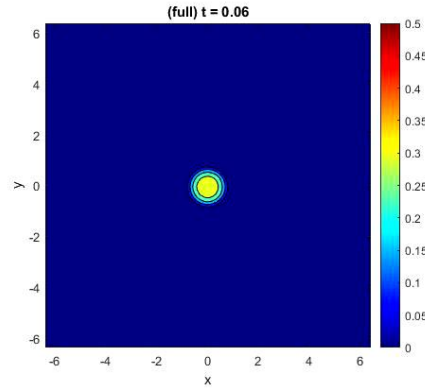
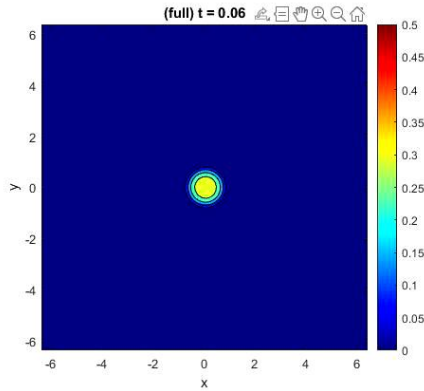


Out-of-sample simulation ($z = +1.7$)

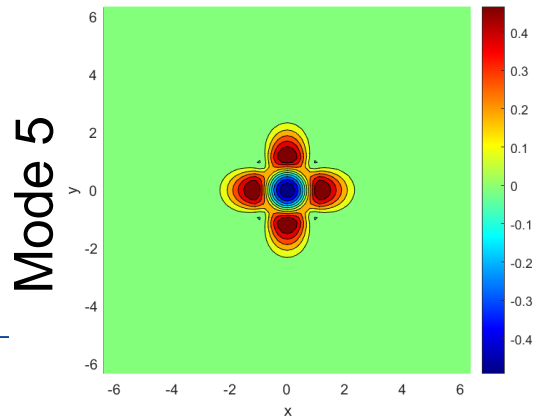
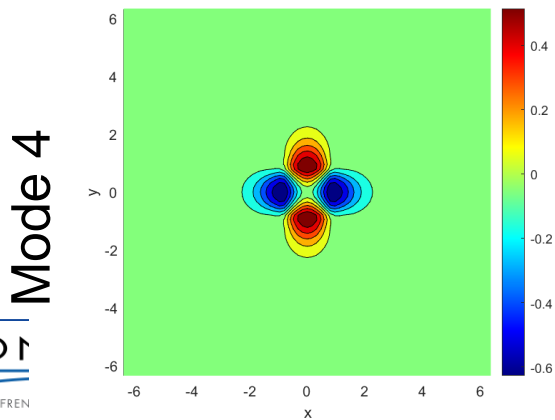
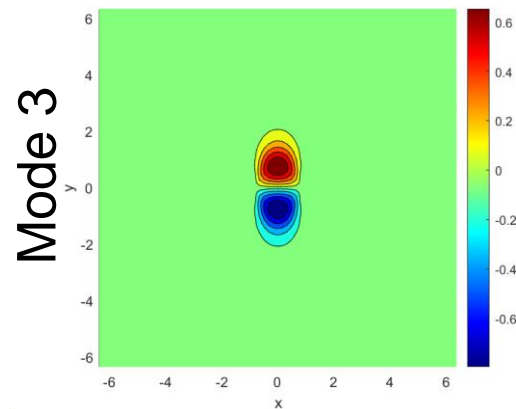
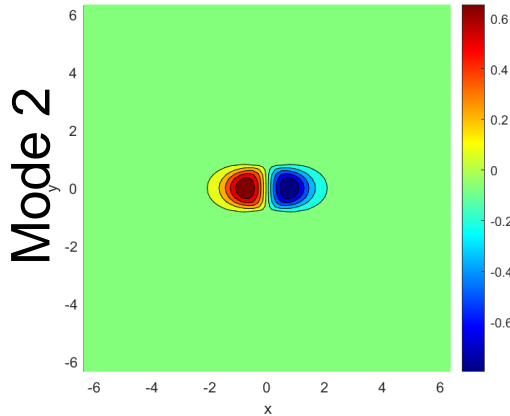
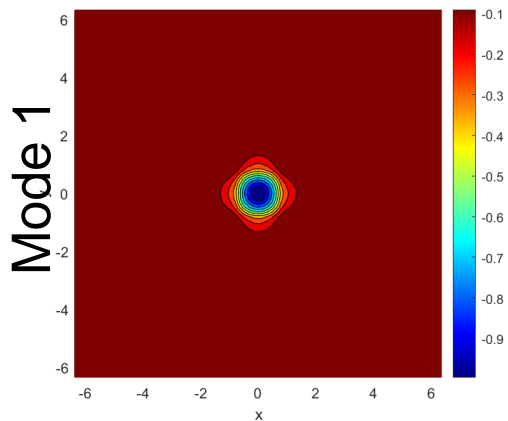


Pure advective problem for a compact support function

$$u(x, t; z) = \exp \frac{-1}{1 - |x - zt|^2} \chi_{\{|x-zt| \leq 1\}}(\mathbf{x})$$

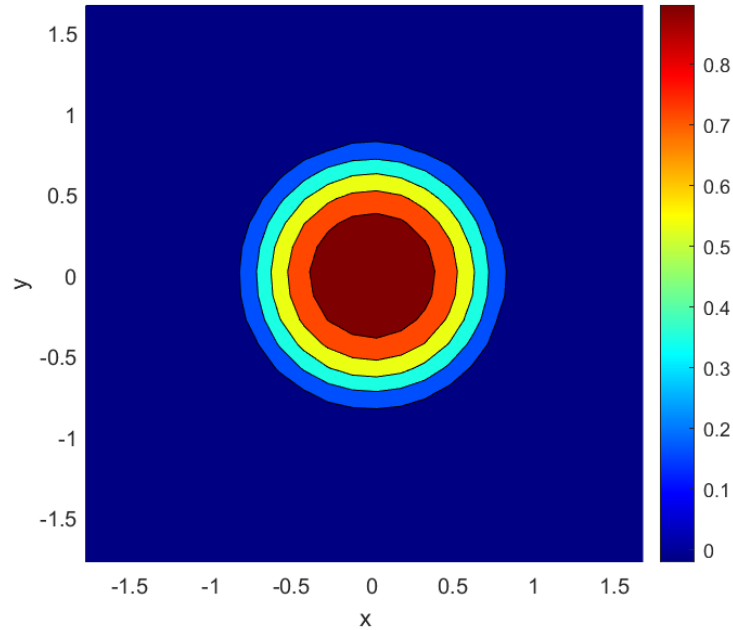


Pure advective problem (one-block approach)

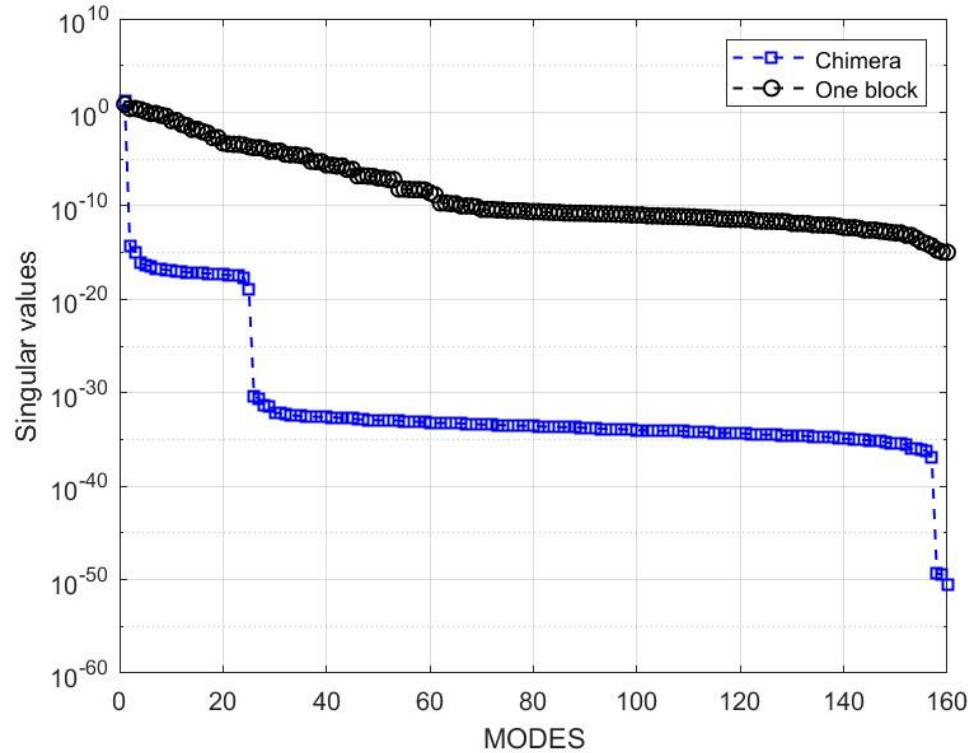


Pure advective problem (Chimera approach)

Mode 1



Comparison of decays of SVs



In-sample and out-of sample simulations

Error Chimera: $\mathcal{O}(10^{-6})$ with 1 mode.

Error one-block: $\mathcal{O}(1)$ with 25 modes.

$$N = 7137; N_{bg} = 6561; N_{fg} = 576;$$
$$N_{fg}^{HF} = 1 \text{ !!!!!!!!!!!!!}$$

Conclusions

- Hyper-reduced ADER-FV
- Collocated method
- Hyper-reduction as mesh-reduction AND extension tool
- Tool able to overcome the Kolmogorov n -width.



Future perspectives

- Extension to (in)compressible NS
- Extension to more classical FV approaches
- Extension to (pure) hyperbolic PDEs (need of good transmission of shocks and discontinuities along interface)



Merci

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