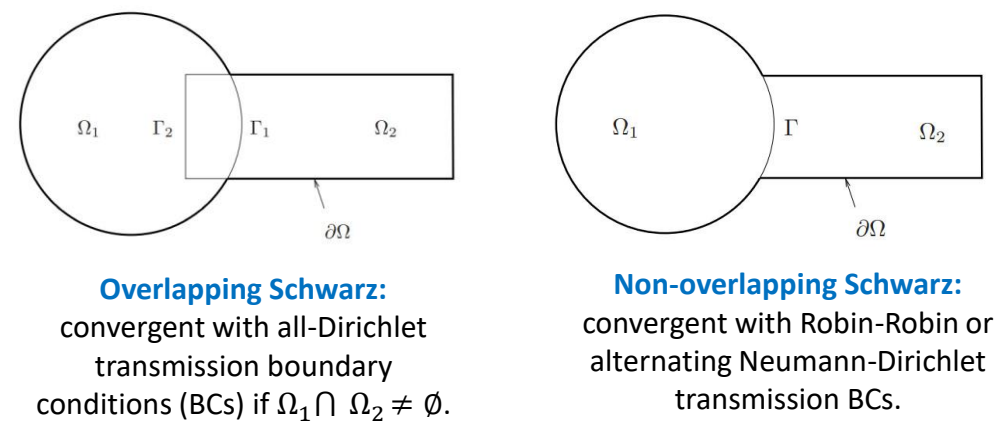




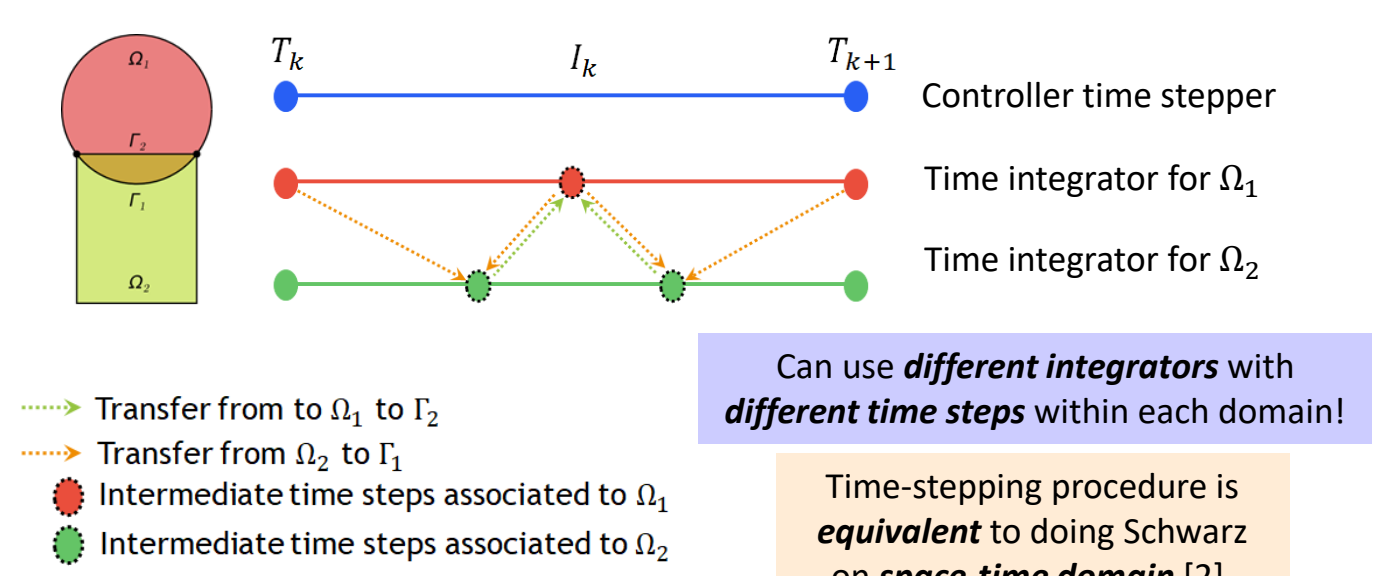
Alternating Schwarz-Based Coupling

Alternating Schwarz Method for Domain Decomposition (DD) and Coupling

- Proposed in 1870 by H. Schwarz for solving Laplace equation on irregular domains.
- Crux of Method:** if the solution of a partial differential equation (PDE) is known in regularly shaped domains, use those as pieces to iteratively build a solution for the more complex domain.
- Basic Schwarz Algorithm for Spatial Coupling**
 - Initialize:** Solve PDE by any method on Ω_1 w/ initial guess for transmission BCs on Γ_1 .
 - Iterate until convergence:**
 - Solve PDE by any method on Ω_2 w/ transmission BCs on Γ_2 based on values just obtained for Ω_1 .
 - Solve PDE by any method on Ω_1 w/ transmission BCs on Γ_1 based on values just obtained for Ω_2 .

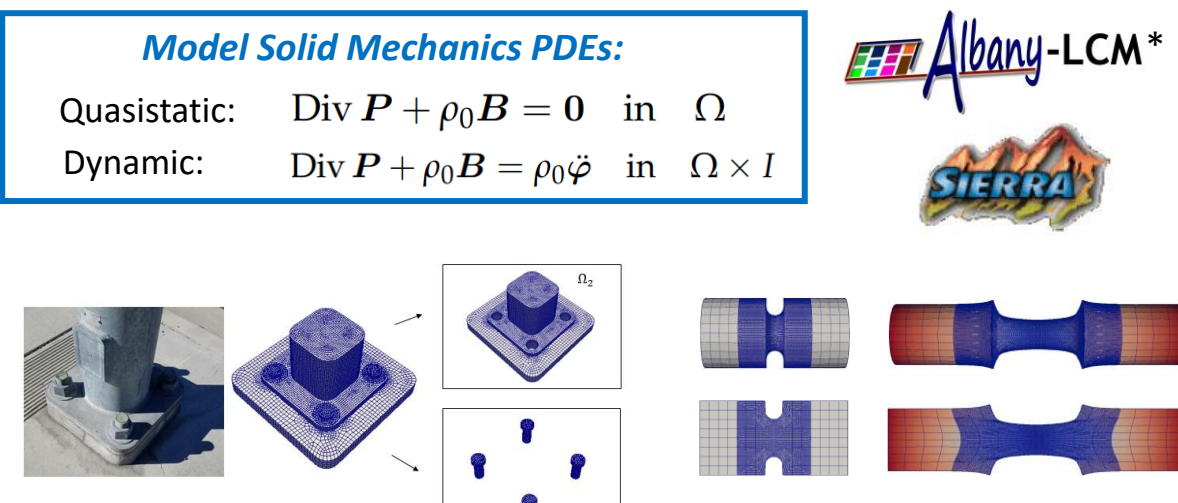


- Novel Idea:** using Schwarz alternating method as a discretization method for solving multi-scale or multi-physics PDEs.
- Schwarz Algorithm for Dynamics**
 - Set $k = 0$ (controller time index).
 - Iterate until convergence:**
 - Step 1:** Advance Ω_1 solution from time T_k to time T_{k+1} using time-stepper in Ω_1 with time-step Δt_1 , using solution in Ω_2 interpolated to Γ_1 at times $T_k + n\Delta t_1$.
 - Step 2:** Advance Ω_2 solution from time T_k to time T_{k+1} using time-stepper in Ω_2 with time-step Δt_2 , using solution in Ω_1 interpolated to Γ_2 at times $T_k + n\Delta t_2$.
 - Step 3:** Check for convergence at time T_{k+1} . If unconverged, return to Step 1. If converged, set $k = k + 1$ and return to Step 1.



Schwarz for Multi-Scale Coupling of Full Order Models (FOMs) in Solid Mechanics

- Coupling is **concurrent** (two-way).
- Easy to implement** into existing massively-parallel HPC codes.
- Scalable, fast, robust** (we target real engineering problems, e.g., analyses involving failure of bolted components).
- Coupling does not introduce **nonphysical artifacts**.
- Theoretical convergence** properties/guarantees [1, 2].
- “Plug-and-play” framework:**
 - Ability to couple regions with **different non-conformal meshes, different element types and different levels of refinement** to simplify task of **meshing complex geometries**.
 - Ability to use **different solvers/time-integrators** in different regions.



Schwarz Extensions to ROM-FOM and ROM-ROM Couplings

- Enforcement of Dirichlet boundary conditions (DBC) in ROM at indices i_{Dir}**
 - Method 1 in [3] is employed: $u(t) \approx \tilde{u} + \Phi \tilde{u}(t)$, $v(t) \approx \tilde{v} + \Phi \tilde{v}(t)$, $a(t) \approx \tilde{a} + \Phi \tilde{a}(t)$.
 - POD modes made to satisfy homogeneous DBCs: $\Phi(i_{Dir}, \cdot) = 0$.
 - BCs imposed by modifying \tilde{u} , \tilde{v} , \tilde{a} : $\tilde{u}(i_{Dir}, \cdot) \leftarrow \chi_u \tilde{u}(i_{Dir}, \cdot) - \chi_v \tilde{v}(i_{Dir}, \cdot) - \chi_a \tilde{a}(i_{Dir}, \cdot)$.
- Choice of domain decomposition**
 - Error-based indicators that help decide in what region of the domain a ROM can be viable should drive domain decomposition (future work).

Snapshot collection and reduced basis construction

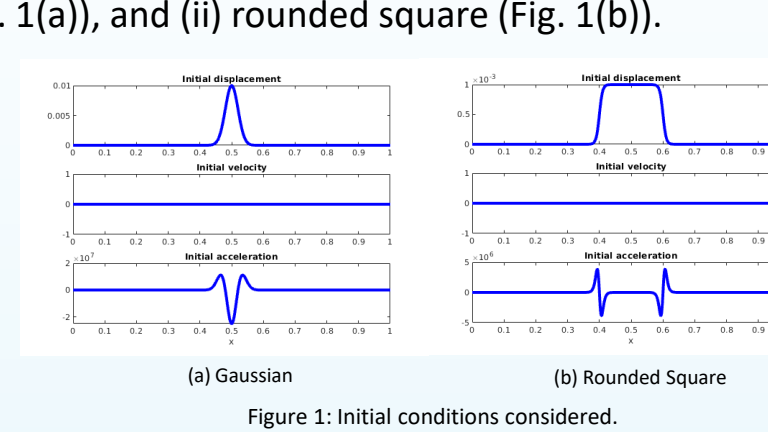
- POD results presented herein use snapshots obtained via FOM-FOM coupling on $\Omega_1 \cup \Omega_2$.
- Future work: generate snapshots/bases separately in each Ω_i .

For nonlinear solid mechanics, hyper-reduction methods need to preserve Hamiltonian structure

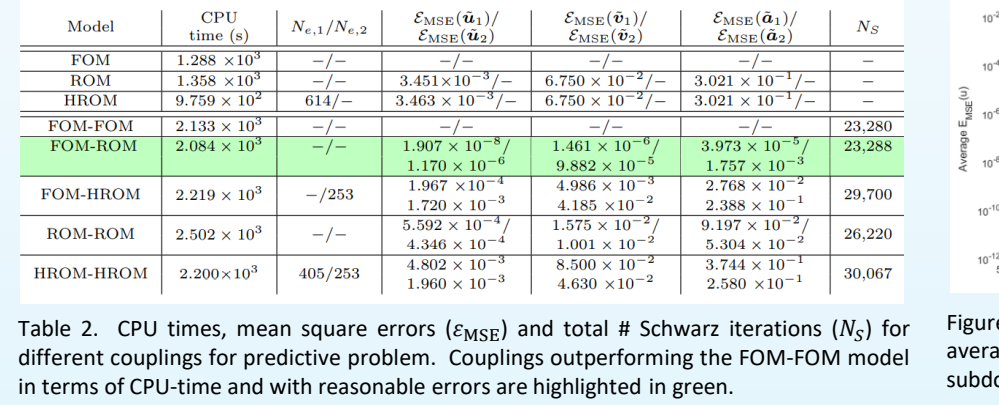
- We employ the Energy-Conserving Sampling & Weighting Method (ECSW).
- Boundary points must be included in sample mesh for DBC enforcement.

Numerical Example: Nonlinear Hyper-Elastic Wave Propagation Problem [4]

- Dynamic solid mechanics** problem with **nonlinear hyperelastic Hencky** constitutive model.
- 1D beam** geometry $\Omega = (0, 1)$, clamped at both ends, with prescribed initial condition discretized using FEM w/ implicit Newmark- β .
- Two initial conditions:** (i) Gaussian (Fig. 1(a)), and (ii) rounded square (Fig. 1(b)).
 - Reproductive case:** generate snapshots with initial condition (i) and predict solution with initial condition (i)
 - Predictive case:** generate snapshots with initial condition (ii) and predict solution with initial condition (i)
- Very stringent test case** for coupling and for ROMs (traveling wave, sharp gradients in solution!)
- Non-overlapping domain decomposition (DD)** of $\Omega = \Omega_1 \cup \Omega_2$, where $\Omega_1 = [0, 0.6]$ and $\Omega_2 = [0.6, 1.0]$ and same $\Delta x = 10^{-3}$ and $\Delta t = 10^{-7}$ in both subdomain.
 - Transmission BCs:** alternating Dirichlet-Neumann with no relaxation.



Model	$M_1(N_1)$	$M_2(N_2)$	CPU time (s)	$E_{err}(u)$	$E_{err}(v)$	$E_{err}(a)$	N_p
FOM-FOM	662	662	1.83	1.00e-07	1.00e-07	1.00e-07	20,000
FOM-ROM	662	662	1.83	1.00e-07	1.00e-07	1.00e-07	20,000
ROM-ROM	662	662	1.83	1.00e-07	1.00e-07	1.00e-07	20,000



Motivation for Coupling

The past decades have seen tremendous investment in simulation frameworks for coupled multi-scale and multi-physics problems.

- Frameworks rely on **established mathematical theories** to couple physics components.
- Most existing coupling frameworks are based on **traditional discretization methods**.

Complex System Model

- PDEs, ODEs
- Nonlocal integral
- Atomistic, ...

Traditional Methods

- Mesh-based (FE, FV, FD)
- Meshless (SPH, MLS)
- Partitioned (loose) coupling
- Eulerian, Lagrangian, ...

Coupled Numerical Model

- Monolithic (Lagrange multipliers)
- Partitioned (loose) coupling
- Iterative (Schwarz, optimization)

Traditional + Data-Driven Methods

- Physics-Informed Neural Networks (PINNs)
- Neural ODEs
- Projection-based ROMs, ...

Unfortunately, existing algorithmic and software infrastructures are **ill-equipped** to handle plug-and-play integration of **data-driven models!**

Objective: discover **mathematical principles** guiding assembly of standard and data-driven models in **stable, accurate and physically consistent ways.**

Projection-Based Model Order Reduction

Full Order Model (FOM): $M \frac{d^2 u}{dt^2} + f_{int}(u) = f_{ext}$

- Acquisition:** Solve ODE at different design points. Save solution data.
- Learning:** Proper Orthogonal Decomposition (POD): $X = U \Sigma V^T$.
- Projection-Based Reduction:** Reduce the number of unknowns. Perform Galerkin projection: $\Phi^T M \Phi \frac{d^2 \tilde{u}}{dt^2} + \Phi^T f_{int}(\Phi \tilde{u}) = \Phi^T f_{ext}$. Hyper-reduce nonlinear terms: $f_{int}(\Phi \tilde{u}) \approx A f_{int}(\tilde{u})$.

ROM = Reduced Order Model
HROM = Hyper-reduced ROM

Coupling methods are **not limited** to linear bases (e.g., POD) and Galerkin projection.

Comparison of Coupling Methods

- Alternating Schwarz-based Coupling**
 - Can do FOM-FOM, FOM-ROM, ROM-ROM coupling
 - Overlapping or non-overlapping DD
 - Iterative formulation (less intrusive but likely requires more CPU time)
 - Can couple different mesh resolutions and element types
 - Can use different time-integrators with different time-steps in different subdomains
 - No interface bases required
 - Sequential subdomain solves in multiplicative Schwarz variant
 - Parallel subdomain solves possible with additive Schwarz variant (not shown)
 - Extensible in straightforward way to PINN/Dynamic Mode Decomposition (DMD) data-driven model
- Lagrange Multiplier-Based IVR Coupling**
 - Can do FOM-FOM, FOM-ROM, ROM-ROM coupling
 - Non-overlapping DD
 - Monolithic formulation requiring hybrid formulation (more intrusive but more efficient)
 - Can couple different mesh resolutions and element types
 - Can use different explicit or IMEX time-integrators with different time-steps in different subdomains
 - Probably convergent variant requires interface bases
 - Parallel subdomain solves if explicit or IMEX time-integrator is employed
 - Extensions to PINN/DMD data-driven models are not obvious

Current & Future Work

- Extension of Schwarz method to multi-D problems.
- Extension of Lagrange multiplier-based method to predictive and nonlinear problems.
- Extension of methods to PINN-PINN coupling.
- Development of error indicator-based DD approaches, to determine optimal ROM-FOM placement (including on-the-fly ROM-FOM switching).
- Extension of couplings to POD modes built from snapshots on independently-simulated subdomains.
- Application to multi-physics problems.
- Journal articles on new methods are in preparation.

References

- [1] A. Mota, I. Tezaur, C. Allenen, "The Schwarz Alternating Method in Solid Mechanics", CMAA, 159 (2017), 19-51.
- [2] A. Mota, I. Tezaur, G. Phipps, "The Schwarz Alternating Method for Dynamic Solid Mechanics", CMAA, 121 (2015), 2003-2017.
- [3] M. Gumbart, J. Peterson, J. Shamba, "Reduced-order modeling of time-dependent PDEs with multiple parameters in the boundary layer", CMAA, 196 (2017), 2020-2047.
- [4] I. Tezaur, A. Mota, "The Schwarz alternating method for the seamless coupling of nonlinear ROMs and FOMs", CMI Summer Proceedings 2022, 296, pp. 31-35.
- [5] A. Mota, P. Bochev, P. Kuberry, "Topology-agnostic partitioned algorithms for interface problems based on Lagrange multipliers", CMAA, 201 (2019), 49-62.
- [6] J.M. Carrero, C. Sockwell, A. Multiscale Discontinuous-Galerkin-in-Time Framework for Interface-Coupled Problems, SAMT, Numer. Anal., 30(8), 273-304, 2022.
- [7] S. Schwab, K. Petersen, P. Kuberry, P. Bochev, Interface Flux Recovery Framework for Constructing Partitioned Heterogeneous Time-Integration Methods, to appear.
- [8] A. de Castro, P. Kuberry, I. Tezaur, P. Bochev, "A synchronous partitioned scheme for coupled reduced order models based on separate reduced order bases for interface and interior variables", CSW Summer Proceedings 2022, 36, pp. 39-52.

Coupling via Generalized Mortar Methods

Lagrange Multiplier-Based Partitioned Coupling Formulation

Model problem: time-dependent advection-diffusion problem on $\Omega = \Omega_1 \cup \Omega_2$ with $\Omega_1 \cap \Omega_2 = \emptyset$

$$\begin{cases} \dot{c}_i - \nabla \cdot F_i(c_i) = f_i, & \text{in } \Omega_i \times [0, T] \\ c_i = g_i, & \text{on } \Gamma_i \times [0, T] \\ c_i(x, 0) = c_{i,0}(x), & \text{in } \Omega_i \end{cases} \quad (1)$$

$i \in \{1, 2\}$, c_i : unknown scalar solution field
 f_i : body force, g_i : boundary data on Γ_i
 $F_i(c_i) := \kappa_i \nabla c_i - \mathbf{u} c_i$: total flux function
 κ_i : non-negative diffusion coefficient
 \mathbf{u} : given advection velocity field

Hybrid semi-discrete coupled formulation: obtained by differentiating interface conditions in time and discretizing hybrid problem using FEM in space

$$\begin{pmatrix} M_1 & 0 & G_1^T \\ 0 & M_2 & -G_2^T \\ G_1 & -G_2 & 0 \end{pmatrix} \begin{pmatrix} \dot{c}_1 \\ \dot{c}_2 \\ \lambda \end{pmatrix} = \begin{pmatrix} f_1 - K_1 c_1 \\ f_2 - K_2 c_2 \\ 0 \end{pmatrix} \quad (2)$$

M_i : mass matrices, K_i : stiffness matrices, G_i : constraint matrices

Decoupling via Schur complement: equation (2) is equivalent to

$$\begin{pmatrix} M_1 & 0 \\ 0 & M_2 - G_2^T G_1^{-1} G_1 \end{pmatrix} \begin{pmatrix} \dot{c}_1 \\ \dot{c}_2 \end{pmatrix} = \begin{pmatrix} f_1 - K_1 c_1 - G_1^T \lambda \\ f_2 - K_2 c_2 + G_2^T \lambda \end{pmatrix} \quad (3)$$

Equations decouple if using explicit or IMEX time-integration!

where $(G_1 M_1^{-1} G_1^T + G_2 M_2^{-1} G_2^T) \lambda = G_1 M_1^{-1} (f_1 - K_1 c_1) - G_2 M_2^{-1} (f_2 - K_2 c_2)$ (4)

- Compatibility conditions:** on interface $\Gamma \times [0, T]$
- Continuity of states:** $c_1(x, t) - c_2(x, t) = 0$
 - Continuity of total flux:** $F_1(x, t) \cdot \mathbf{n}_1 = F_2(x, t) \cdot \mathbf{n}_1$
 - \Rightarrow Imposed weakly using Lagrange multiplier (LM) λ
- Implicit Value Recovery (IVR) Algorithm [5]**
- Pick explicit or IMEX time-integration scheme for Ω_1 and Ω_2 .
 - Time integration schemes and time-steps in Ω_1 and Ω_2 can be different!
 - Approximate LM space as trace of finite element space on Ω_1 or Ω_2 .
 - Compute matrices M_i , K_i , G_i and vectors f_i .
 - For each timestep t^n :
 - Solve Schur complement system (4) for the LM λ^n
 - Update the state variables c_i^n by advancing (3) in time
- * Ensures that dual Schur complement of (2) is symmetric positive definite

Lagrange Multiplier-Based FOM-ROM and ROM-ROM Coupling

- Collect snapshots using suitable monolithic FOM solve for equation (1) and subtract Dirichlet BC data on $\Gamma_1 \cup \Gamma_2$.
 - Partition modified snapshots into subdomain snapshot matrices X_1 and X_2 on Ω_1 and Ω_2 , respectively.
 - Calculate "split" reduced bases $\Phi_{1,r}$ and $\Phi_{2,r}$, for interface and interior degrees of freedom (DOFs).
 - Approximate the solution as a linear combination of the POD modes in each subdomain and on each boundary: $c_{i,0}(t) \approx \tilde{c}_{i,0}(t) + \Phi_{i,0} \tilde{e}_{i,0}(t)$, $c_{i,r}(t) \approx \tilde{c}_{i,r}(t) + \Phi_{i,r} \tilde{e}_{i,r}(t)$
 - Reduce LM space to size $N_{r,LM} < N_{r,1,r} + N_{r,2,r}$, where $N_{r,i,r} = \#$ POD modes in $\Phi_{i,r}$, and approximate $\lambda \approx \Phi_{LM} \tilde{\lambda}$ where $\Phi_{LM} = \Phi_{1,r}$ for $i = 1, 2$, so that $N_{r,LM} = N_{r,1,r}$.
 - Substitute above expansions into (2) and project equations onto reduced bases to obtain:
$$\begin{pmatrix} \tilde{M}_{1,r} & \tilde{M}_{1,r,0} & 0 & 0 & \tilde{G}_{1,r}^T \\ \tilde{M}_{1,0,r} & \tilde{M}_{1,0,0} & 0 & 0 & 0 \\ 0 & 0 & \tilde{M}_{2,r,0} & \tilde{M}_{2,r,0} & -\tilde{G}_{2,r}^T \\ 0 & 0 & \tilde{M}_{2,0,r} & \tilde{M}_{2,0,0} & 0 \\ \tilde{G}_{1,r} & 0 & -\tilde{G}_{2,r} & 0 & \lambda \end{pmatrix} \begin{pmatrix} \dot{\tilde{c}}_{1,r} \\ \dot{\tilde{c}}_{1,0} \\ \dot{\tilde{c}}_{2,r} \\ \dot{\tilde{c}}_{2,0} \\ \lambda \end{pmatrix} = \begin{pmatrix} s_{1,r} \\ s_{1,0} \\ s_{2,r} \\ s_{2,0} \\ 0 \end{pmatrix}$$
- Extensions to FOM-ROM coupling is straight-forward.

Numerical Example: High-Peclet Advection-Diffusion Problem [8]

Errors for ROM-ROM coupling with full and reduced LM spaces are identical to machine precision (Fig. 11).

- As expected, using reduced LM space improves condition number (Fig. 12).
- Conditioning of the Schur complement for ROM-ROM with reduced LM space is essentially the same as for the FOM-FOM coupling (proven to be well-conditioned in [5]) (Fig. 12).
- Each coupling is capable of attaining an error on the order of the relative error for the FOM-FOM coupling (Fig. 13).
- Reduced LM ROM-ROM coupling achieves optimal errors in less time (Fig. 13).
- Couplings deliver solutions which are smooth and artifact-free (Fig. 14).

Key Takeaways

- All coupled models evaluated converge in ≤ 3 Schwarz iterations/- FOM-FOM and HROM-FOM couplings outperform FOM-FOM coupling in CPU time by 12.5-32.6%. Greater speedups expected in multi-D.
- All couplings involving ROMs/HROMs are at least as accurate as single-domain ROMs/HROMs.
- Predictive couplings involving ROMs/HROMs are smooth and oscillation-free!