



# NON-INTRUSIVE COUPLING OF REDUCED-ORDER AND FULL-ORDER MODELS WITH ADAPTIVE DOMAIN DECOMPOSITION

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## MOTIVATION for ROM-FOM coupling

Reduced-order models [ROMs] achieve significant computational speed-up through projection onto low-dimensional reduced spaces but are only as expressive as their reduced spaces. Meanwhile, full-order models [FOMs] are highly expressive but computationally slow.

**Objective:** Attain computational efficiency at the target accuracy through adaptive, local ROM-FOM coupling.

## RESEARCH GOALS

### ROM-FOM coupling with adaptive subdomains:

- Use ROM for computational speed-up
- Use *local* FOM for under-resolved features
- Retrain ROM on generated FOM data online to improve resolution
- Adjust the FOM subdomain for comp. speed-up

### Non-intrusive ROMs for modular components:

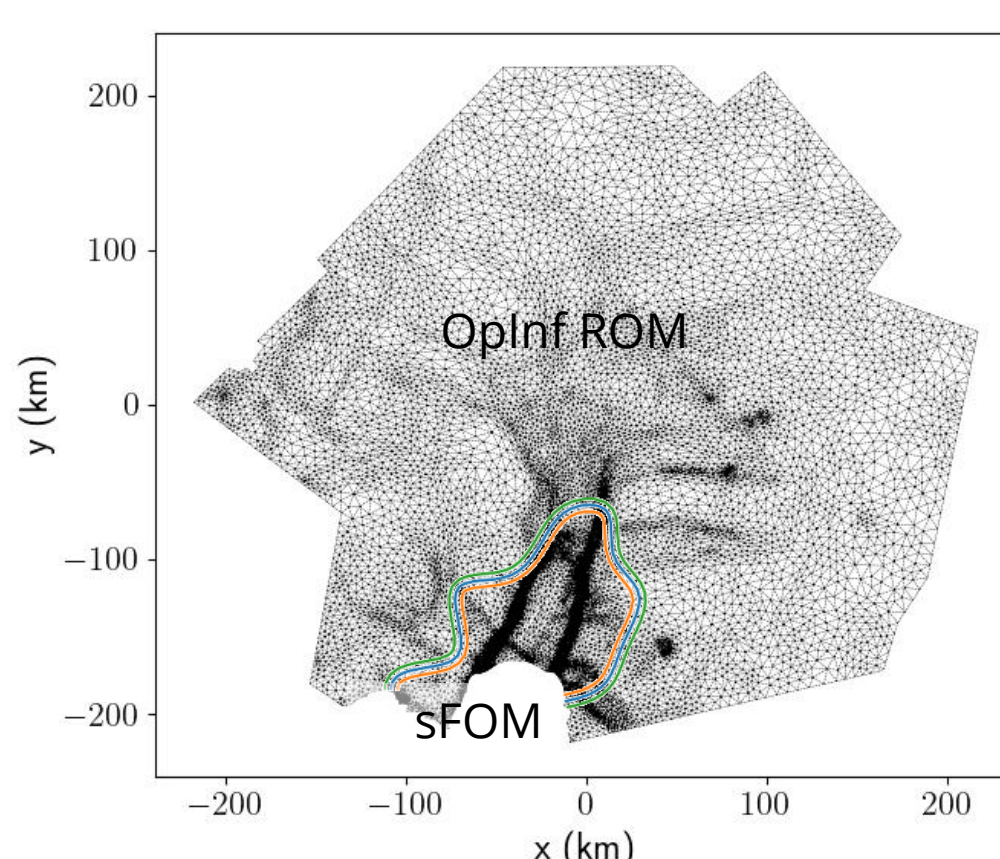
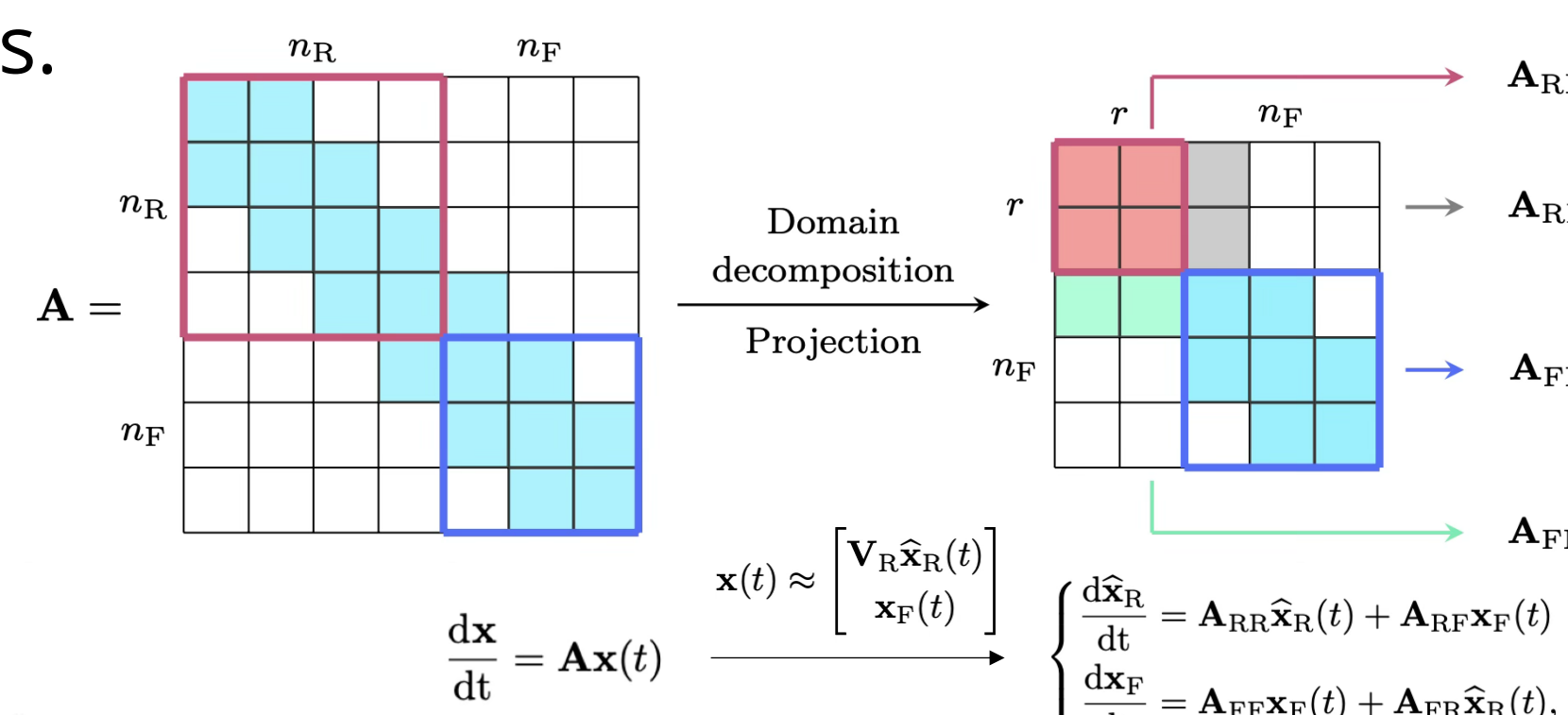
- Offline: Train bubble-ROM for reference components
  - Train on individual or assembled systems
- Online: Assemble complex geometry via ports
- Keep everything non-intrusive
- Provide error guarantees if possible

## NON-INTRUSIVE COUPLING OF OPINF AND SFOM MODELS

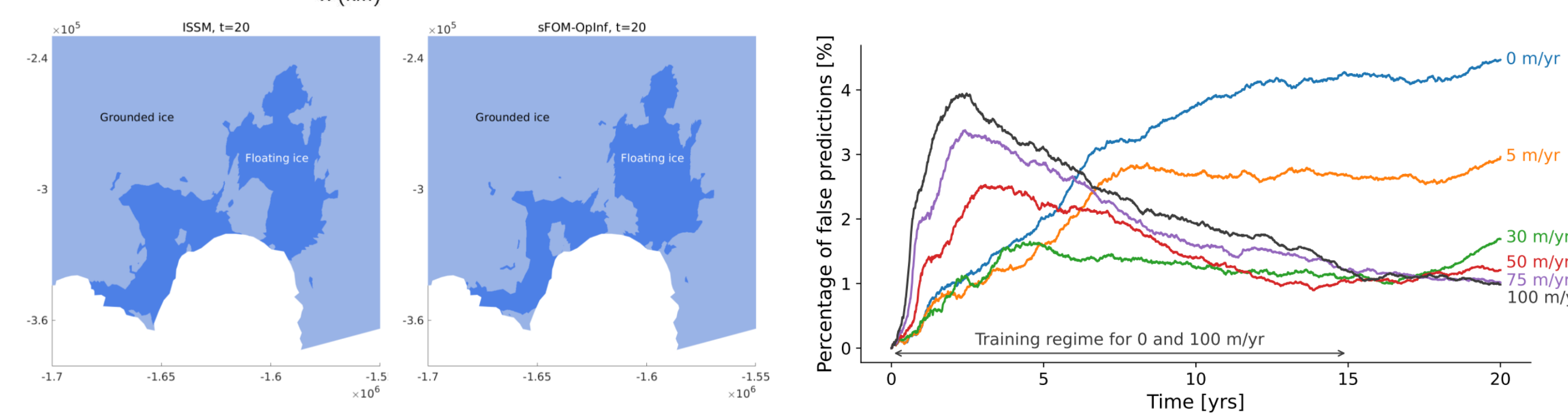
**Operator Inference** [OpInf, 1] learns a ROM on a chosen low-dimensional subspace by leveraging the structure of the governing equations.

**Sparse FOM inference** [sFOM, 2] imposes a sparsity pattern to learn full-order dynamics.

In [3] we combine both approaches to approximate dynamical systems with spatially localized features.



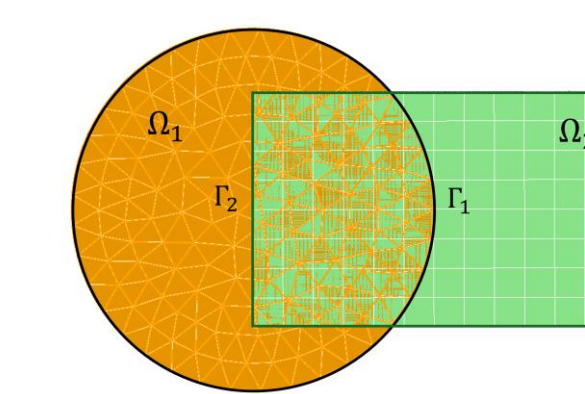
Applied to an ice-thickness model of the Pine Island Glacier in Western Antarctica with varying melt parameters, we achieve below 1% error at 8 x speed-up. Moreover, we accurately predict the grounding line between floating and grounded ice.



## SCHWARZ ALTERNATING METHOD FOR OPINF COUPLING

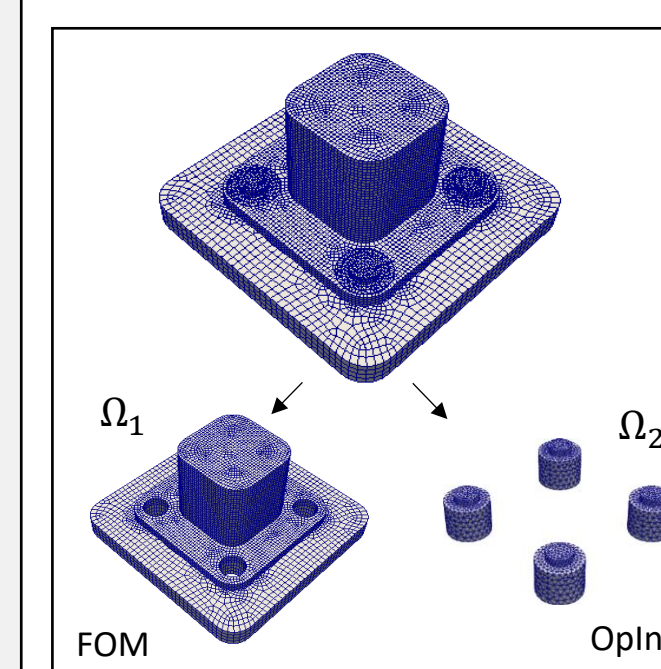
### Basic Schwarz algorithm for coupling

- Decompose  $\Omega = \Omega_1 \cup \Omega_2$
- Solve PDE by any method in  $\Omega_2$  w/ BCs on  $\Gamma_2$  from just-obtained  $\Omega_1$  solution.
- Solve PDE by any method in  $\Omega_1$  w/ BCs on  $\Gamma_1$  from just-obtained  $\Omega_2$  solution.
- Iterate until convergence

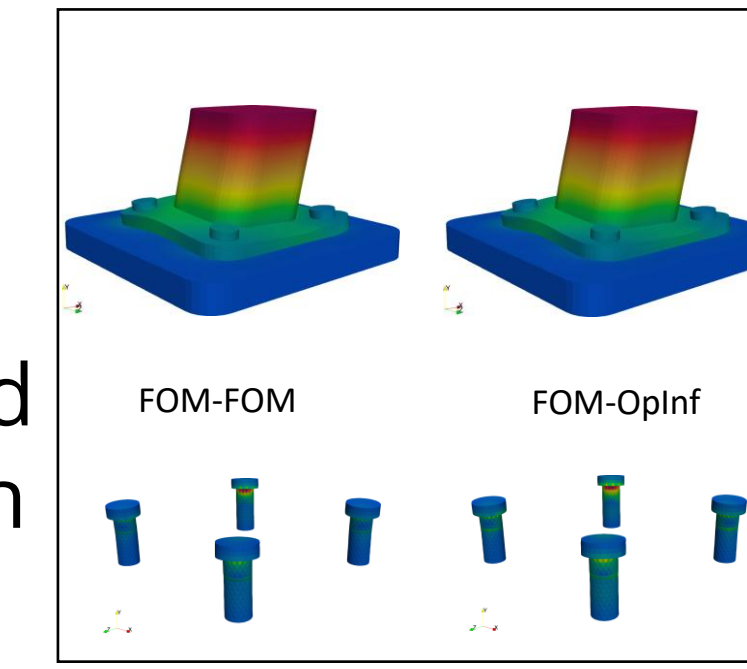


OpInf ROM + Schwarz BCs in  $\Omega_i$ :  $\hat{q}_i + \hat{A}_i \hat{q}_i + \hat{H}_i(\hat{q}_i \otimes \hat{q}_i) + \hat{B}_i g_i = 0$   
 OpInf BC implementation is motivated Dirichlet BC treatment in FEM  
 Learned Schwarz Dirichlet BC term

Schwarz enables the seamless coupling of disparate models (FOM, OpInf), meshes, solvers and time-steppers.



Schwarz has achieved meshing simplification & successful FOM-OpInf coupling for complex solid mechanics problems, with speedups up to 106x [4].



## REFERENCES

- [1] Peherstorfer, Willcox (2016). Data-driven operator inference for nonintrinsic projection-based model reduction.
- [2] Gkimitis, Richter, Benner (2024). Adjacency-based, non-intrinsic model reduction for vortex-induced vibrations.
- [3] Gkimitis, Aretz, Tezzele, Richter, Benner, Willcox (2025). Non-intrinsic reduced-order modeling for dynamical systems with spatially localized features.
- [4] Tezaur, Parish, Gruber, Moore, Wentland, Mota (2026, submitted). Hybrid coupling with operator inference and the Schwarz alternating method.