

Two-Scale Reduction of LOD Multiscale Models

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Model Reduction and Surrogate Modeling (MORE)

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Outline

1. LOD

2. RBLOD

3. TS LOD

4. TSRBLOD

where

- ▶ LOD = Localized Orthogonal Decomposition
- ▶ RB = Reduced Basis
- ▶ TS = Two-Scale

Multiscale Model Problem

Parameterized diffusion equation

For a fixed parameter $\mu \in \mathcal{P}$ find u_μ s.t.

$$\begin{aligned} -\nabla \cdot A_\mu \nabla u_\mu &= f, & \text{in } \Omega, \\ u_\mu &= 0, & \text{on } \partial\Omega, \end{aligned}$$

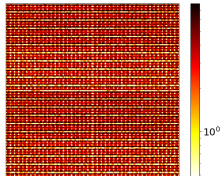
or in weak form

$$a_\mu(u_\mu, v) = F(v), \quad \forall v \in V$$

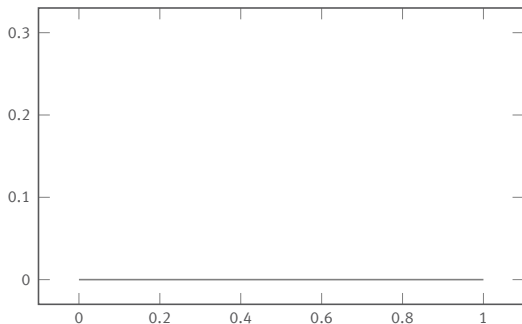
- ▶ Parameter space $\mathcal{P} \subset \mathbb{R}^p, p \in \mathbb{N}$.
- ▶ Bounded Lipschitz domain $\Omega \subset \mathbb{R}^d$.
- ▶ Homogeneous Dirichlet boundary conditions, $V := H_0^1(\Omega)$.
- ▶ $A_\mu \in L^\infty(\Omega, \mathbb{R}^{d \times d})$ symmetric and uniformly elliptic:

$$0 < \alpha \leq A_\mu \leq \beta < \infty.$$

- ▶ $f \in L^2(\Omega)$.
- ▶ Rapid change of $A_\mu(x)$, but not too high contrast $\kappa := \beta/\alpha$.



Need for Numerical Multiscale Methods

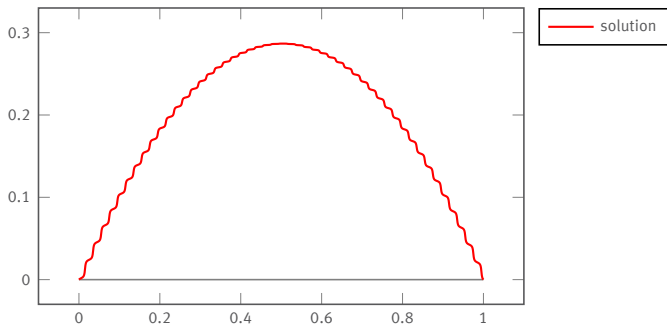


Toy example: For $\varepsilon \ll 1$, find $u_\varepsilon \in H_0^1((0, 1))$ s.t.

$$-\left(A\left(\frac{x}{\varepsilon}\right) u_\varepsilon'(x)\right)' = 1$$

$$A(y) = 1 + 0.9 \sin(2\pi y).$$

Need for Numerical Multiscale Methods

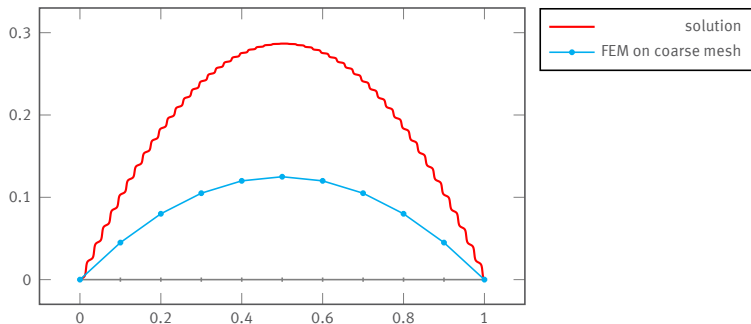


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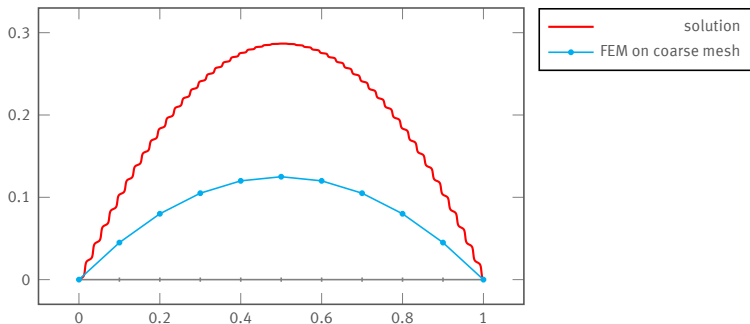


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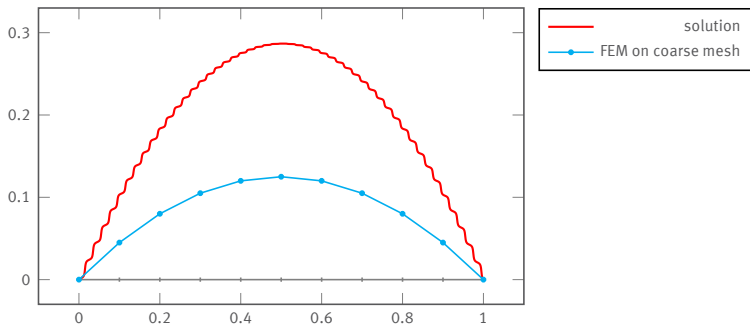
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Need for Numerical Multiscale Methods



Problem: Coarse-mesh FEM space V_H is a bad H^1 -approximation space for u_ϵ

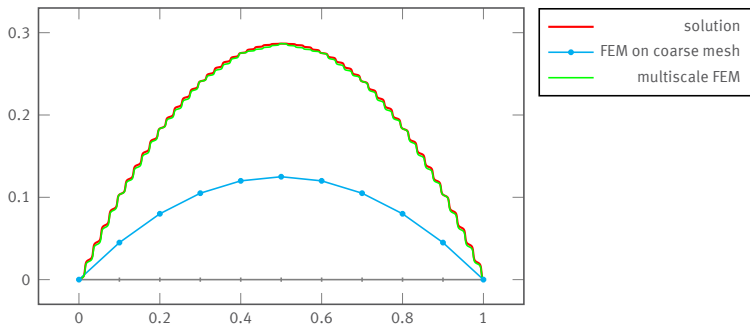
Need for Numerical Multiscale Methods



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Solution: Find *multiscale* FEM space V_H^{ms} .

Need for Numerical Multiscale Methods

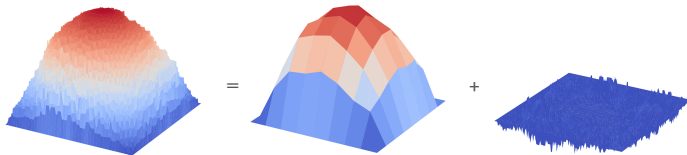


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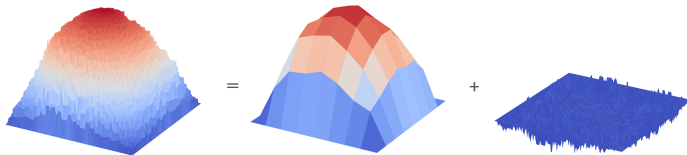
Multiscale Orthogonal Decomposition

- ▶ Fine mesh \mathcal{T}_h and coarse mesh \mathcal{T}_H with maximal element diameter $H \gg h$, FE spaces V_h and $V_H := V_h \cap \mathcal{P}_1(\mathcal{T}_H)$.
- ▶ Quasi-interpolation operator $\mathcal{J}_H: V_h \rightarrow V_H$ (e.g. based on local L^2 -projections).
- ▶ Fine-scale space $V^f := \ker(\mathcal{J}_H) \rightsquigarrow$ decomposition $V_h = V_H + V^f$.



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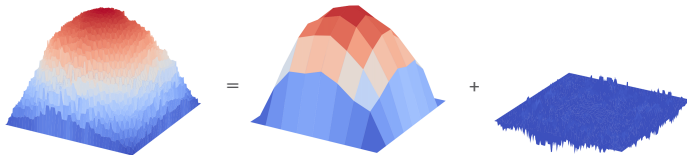
- ▶ Fine-scale correction $\mathcal{Q}_\mu: V_H \rightarrow V^f$ defined by

$$a_\mu(\mathcal{Q}_\mu(v_H), v^f) = a_\mu(v_H, v^f), \quad \forall v^f \in V^f.$$

- ▶ Multiscale space $V_{H,\mu}^{ms} = (I - \mathcal{Q}_\mu)V_H \rightsquigarrow a_\mu$ -orthogonal decomposition $V_h = V_\mu^{ms} \oplus_{a_\mu} V^f$.

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- ▶ Fine-scale correction $Q_\mu: V_H \rightarrow V^f$ defined by

$$a_\mu(Q_\mu(v_H), v^f) = a_\mu(v_H, v^f), \quad \forall v^f \in V^f.$$

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Problem: Computing Q_μ as hard as finding $u_{h,\mu}$!

Localization

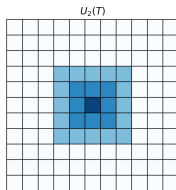
▶ Truncated fine-scale space $V_{h,k,T}^f := \left\{ v \in V^f \mid v|_{\Omega \setminus U_k(T)} = 0 \right\}$.

▶ For each $T \in \mathcal{T}_H$, define localized correctors $Q_{k,\mu}^T(v_H) \in V_{h,k,T}^f$

$$a_\mu(Q_{k,\mu}^T(v_H), v^f) = a_\mu^T(v_H, v^f), \quad \forall v^f \in V_{h,k,T}^f,$$

▶ Localized corrector operator $Q_{k,\mu} = \sum_{T \in \mathcal{T}_H} Q_{k,\mu}^T$.

▶ LOD space $V_{H,k,\mu}^{ms} := (I - Q_{k,\mu})V_H = \{ \phi_x - Q_{k,\mu}(\phi_x) \mid x \in \mathcal{N}_H \}$



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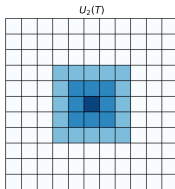
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Lemma [Målqvist/Peterseim '14]

The correctors $Q_\mu(\phi_x)$ decay exponentially, hence,

$$\| Q_\mu(\phi_x) - Q_{k,\mu}(\phi_x) \| \leq C_Q k^{d/2} \theta^k \| Q_\mu(\phi_x) \|,$$

where $0 < \theta < 1$ and C_Q depends on κ but not on the variations of A_μ .

Petrov–Galerkin Formulation [Elfverson/Ginting/Henning '15]

Petrov–Galerkin LOD method

Find $u_{H,k,\mu}^{ms} \in V_{H,k,\mu}^{ms}$ such that

$$a_{\mu}(u_{H,k,\mu}^{ms}, v) = F(v), \quad \forall v \in V_H.$$

Advantages of PG over Galerkin:

- ▶ No coupling between correctors.
- ▶ Reduced memory consumption.
- ▶ Still similar convergence results.

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Convergence theorem

$$\|u_{h,\mu} - u_{H,k,\mu}\|_{L^2} + \|u_{h,\mu} - u_{H,k,\mu}^{ms}\|_1 \lesssim (H + \theta^k k^{d/2}) \|f\|_{L^2(\Omega)}$$

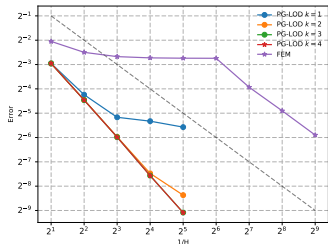


Figure: Energy error $\|u - u_{H,k,\mu}^{ms}\|$ for the PG–LOD and $\|u - u_H\|$ for 1d model problem from [Peterseim'16].

RBLOD¹

[Abdulle, Henning, 15]

Idea: Accelerate LOD by applying RB methodology to corrector problems

$$a(Q_{k,\mu}^T(v_H), v^f) = a_\mu^T(v_H, v^f), \quad \forall v^f \in V_{h,k,T}^f,$$

¹We present here a slight variation of the original work.

RBL0D¹

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$$a(Q_{k,\mu}^T(v_H), v^f) = a_\mu^T(v_H, v^f), \quad \forall v^f \in V_{h,k,T}^f,$$

The approximate LOD coarse-scale system matrix \mathbb{K}_μ^{rb} can then be pre-assembled as

$$(\mathbb{K}_\mu^{rb})_{ji} = a_\mu(\varphi_i - Q_{k,\mu}^{T,rb}(\varphi_i), \varphi_j) = a_\mu(\varphi_i, \varphi_j) - \sum_T a_\mu \left(\underbrace{Q_{k,\mu}^{T,rb}(\varphi_i)}_{\sum_{q=1}^Q \theta_q(\mu) a^q \sum_{n=1}^{N_T} c_n \psi_n^T \in V_{k,T}^{f,rb}}, \varphi_j \right) \quad (*)$$

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Benefit: \mathbb{K}_μ^{rb} can be assembled independently of $\dim V_h$!

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Shortcomings:

- ▶ Size of \mathbb{K}_μ^{rb} still depends on $\dim V_H$, which might be large.
- ▶ Rigorous a posteriori error control for RBLOD solution?
- ▶ Apply RB to coarse-scale problem? (*) leads to large $\dim V_H$ -dependent affine decomposition!

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Two-Scale Formulation of the LOD in homogenization: [Allaire, 92], for HMM: [Ohlberger, 05]

Two-scale space

$$\mathfrak{V} = V_H \oplus V_{h,k,T_1}^f \oplus \cdots \oplus V_{h,k,T_{|\mathcal{T}_H|}}^f$$
$$\|u\|_1^2 = \|u_H\|_1^2 + \sum_{T \in \mathcal{T}_H} \|u_T^f\|_1^2$$

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Two-scale bilinear form

$$\mathfrak{B}_\mu(u, v) := a_\mu(u_H - \sum_{T \in \mathcal{T}_H} u_T^f, v_H) + \rho^{1/2} \sum_{T \in \mathcal{T}_H} a_\mu(u_T^f, v_T^f) - a_\mu^T(u_H, v_T^f),$$

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Proposition

The two-scale solution $u_\mu \in \mathfrak{V}$ of

$$\mathfrak{B}_\mu(u_\mu, v) = F(v_H) \quad \forall v \in \mathfrak{V}.$$

is uniquely determined and given by $u_\mu = [u_{H,k,\mu}, Q_{k,\mu}^{T_1}(u_{H,k,\mu}), \dots, Q_{k,\mu}^{T_{|\mathcal{T}_H|}}(u_{H,k,\mu})]$.

Two-Scale Stability Estimate

Proposition

Let $\rho := C_{\text{ovl}} \cdot \kappa$, then \mathfrak{B}_μ is $\|\cdot\|_{a,\mu} \cdot \|\cdot\|_1$ -continuous and inf-sup stable with the following bounds on the respective constants:

$$\sup_{0 \neq u \in \mathfrak{U}} \sup_{0 \neq v \in \mathfrak{V}} \frac{\mathfrak{B}_\mu(u, v)}{\|u\|_{a,\mu} \cdot \|v\|_1} \leq \beta^{1/2} \quad \text{and} \quad \inf_{0 \neq u \in \mathfrak{U}} \sup_{0 \neq v \in \mathfrak{V}} \frac{\mathfrak{B}_\mu(u, v)}{\|u\|_{a,\mu} \cdot \|v\|_1} \geq \gamma_k / \sqrt{5}.$$

where γ_k is the PG-LOD inf-sup constant and

$$\|u\|_{a,\mu}^2 := \|u_H\|^2 - \sum_{T \in \mathcal{T}_H} u_T^f \|a_{a,\mu}\|_{a,\mu}^2 + \rho \sum_{T \in \mathcal{T}_H} \|Q_{k,\mu}^T(u_H) - u_T^f\|_{a,\mu}^2$$

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Error Bound

$$\|u_\mu - u\|_{a,\mu} \leq \sqrt{5} \gamma_k^{-1} \sup_{v \in \mathfrak{V}} \frac{\mathfrak{F}(v) - \mathfrak{B}_\mu(u, v)}{\|v\|_1} \leq \sqrt{5} \underbrace{\gamma_k^{-1} \beta^{1/2}}_{\sim \kappa^{1/2}} \|u_\mu - u\|_{a,\mu}.$$

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Error Bound

$$\begin{aligned} & \left\{ \|u_{H,k,\mu} - u_H\|_1^2 + \rho \sum_{T \in \mathcal{T}_H} \|Q_{k,\mu}^T(u_H) - u_T^f\|_1^2 \right\}^{1/2} \leq \sqrt{5} C_{\mathcal{J}_H} \alpha^{-1/2} \gamma_k^{-1} \sup_{v \in \mathfrak{D}} \frac{\mathfrak{F}(v) - \mathfrak{B}_\mu(u, v)}{\|v\|_1} \\ & \leq \sqrt{15} C_{\mathcal{J}_H} (C_{\text{ovl}} + 1)^{1/2} \underbrace{\kappa^{1/2} \gamma_k^{-1} \beta^{1/2}}_{\sim \kappa} \left\{ \|u_{H,k,\mu} - u_H\|_1^2 + \rho \sum_{T \in \mathcal{T}_H} \|Q_{k,\mu}^T(u_H) - u_T^f\|_1^2 \right\}^{1/2}. \end{aligned}$$

Two-Scale Reduced Basis Approach

Idea: Accelerate LOD by applying RB methodology to two-scale problem

$$\mathfrak{B}_\mu(u_\mu, v) = F(v_H) \quad \forall v \in \mathcal{V}.$$

Since we only have inf-sup stability of \mathfrak{B}_μ , use as ROM:

$$u_\mu^{rb} := \arg \min_{u \in \mathcal{U}^{rb}} \sup_{v \in \mathcal{V}} \frac{\mathfrak{F}(v) - \mathfrak{B}_\mu(u, v)}{\|v\|_1}.$$

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Benefits:

- ▶ ROM independent from $\dim V_h$ and $\dim V_H$.
- ▶ \mathfrak{B}_μ has same affine decomposition as a_μ .
- ▶ Rigorous a posteriori error bounds.

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Snapshots: LOD solutions with correctors.

Caveat: Need to solve fine-scale corrector problems in each greedy iteration.

Two-Stage Two-Scale Reduced Basis Approach

Stage 1 (for each $T \in \mathcal{T}_H$)

FOM:

$$a_\mu(Q_{k,\mu}^T(v_H), v_T^f) = a_\mu^T(v_H, v_T^f), \quad \forall v_T^f \in V_{h,k,T}^f.$$

ROM:

$$a_\mu(Q_{k,\mu}^{T,rb}(v_H), v_T^f) = a_\mu^T(v_H, v_T^f), \quad \forall v_T^f \in V_{k,T}^{f,rb}.$$

Outputs:

$$a_\mu(Q_{k,\mu}^{T,rb}(v_H), \phi_j), \quad a_\mu^T(\phi_j, Q_{k,\mu}^{T,rb}(v_H))$$

Error bound:

$$\|Q_{k,\mu}^T(v_H) - Q_{k,\mu}^{T,rb}(v_H)\|_{a,\mu} \leq \alpha^{-1/2} \sup_{v_T^f \in V_{h,k,T}^f} \frac{a_\mu^T(v_H, v_T^f) - a_\mu(Q_{k,\mu}^{T,rb}(v_H), v_T^f)}{\|v_T^f\|_1}.$$

Basis generation: weak greedy algorithm w. error tolerance ε_1

Two-Stage Two-Scale Reduced Basis Approach

Stage 2

FOM:

$$\mathfrak{B}_\mu(u_\mu, v) = F(v_H) \quad \forall v \in \mathcal{V}.$$

ROM:

$$u_\mu^{rb} := \arg \min_{u \in \mathcal{U}^{rb}} \sup_{v \in \mathcal{V}} \frac{\mathfrak{F}(v) - \mathfrak{B}_\mu(u, v)}{\|v\|_1}.$$

Error bound:

$$\left\{ \|u_{H,k,\mu} - u_H\|_1^2 + \rho \sum_{T \in \mathcal{T}_H} \|Q_{k,\mu}^T(u_H) - u_T^f\|_1^2 \right\}^{1/2} \leq \sqrt{5} C_{\mathcal{J}_H} \alpha^{-1/2} \gamma_k^{-1} \sup_{v \in \mathcal{V}} \frac{\mathfrak{F}(v) - \mathfrak{B}_\mu(u, v)}{\|v\|_1}$$

Basis generation: ‘weak’ greedy algorithm w. error tolerance ε_2

Snapshots:

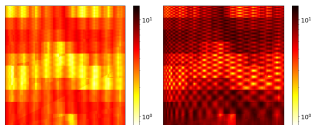
$$K_\mu^{rb} \cdot \underline{u}_{H,k,\mu^*} = F$$

$$u_{\mu^*} := [u_{H,k,\mu^*}, Q_{k,\mu^*}^{T_1,rb}(u_{H,k,\mu^*}), \dots, Q_{k,\mu^*}^{T_r,rb}(u_{H,k,\mu^*})]$$

Numerical Experiment 1

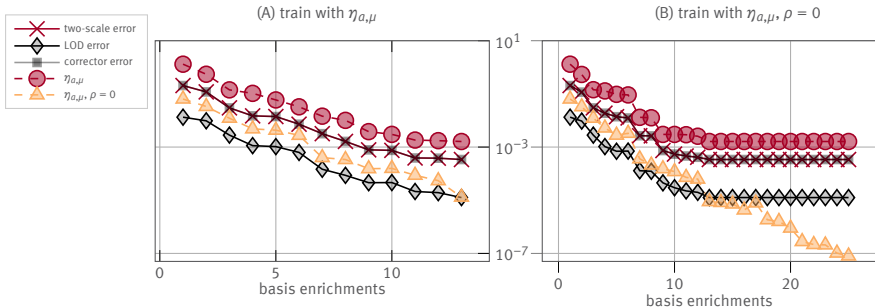
[Abdulle/Henning'15]

- ▶ $\mathcal{P} := [0, 5]$
- ▶ $A_\mu = \sum_{q=1}^Q \theta_q(\mu) A_q$ with $Q = 4$
- ▶ $|\mathcal{T}_h| = 65, 536$
- ▶ $\varepsilon_1 = \varepsilon_2 = 0.001$
- ▶ two samples of A_μ :



| $ \mathcal{T}_h $ | $2^3 \times 2^3$ | | $2^4 \times 2^4$ | | $2^5 \times 2^5$ | |
|-------------------------|------------------|-------------|------------------|-------------|------------------|--------------|
| method | RBL0D | TSRBLOD | RBL0D | TSRBLOD | RBL0D | TSRBLOD |
| $t_{1,av}^{offline}(T)$ | 41 | 61 | 39 | 61 | 33 | 55 |
| $t_1^{offline}$ | 71 | 106 | 67 | 102 | 63 | 98 |
| $t_2^{offline}$ | - | 8 | - | 56 | - | 472 |
| $t^{offline}$ | 71 | 114 | 67 | 158 | 63 | 570 |
| cum. size St.1 | 2346 | 1670 | 8718 | 6134 | 31810 | 22189 |
| av. size St.1 | 9.16 | 26.09 | 8.51 | 23.96 | 7.77 | 21.67 |
| size St.2 | - | 8 | - | 9 | - | 9 |
| t_{LOD} | 0.69 | | 0.49 | | 0.90 | |
| t_{online} | 0.0610 | 0.0003 | 0.2272 | 0.0003 | 1.0462 | 0.0003 |
| speed-up LOD | 11 | 2506 | 2 | 1536 | 1 | 2714 |
| $e_{LOD}^{H^1,rel}$ | 1.97e-5 | 7.30e-4 | 5.08e-5 | 2.94e-4 | 1.11e-4 | 4.21e-4 |
| $e_{LOD}^{L^2,rel}$ | 4.89e-6 | 2.71e-4 | 6.77e-6 | 1.03e-4 | 7.70e-6 | 1.32e-4 |
| $e_{FEM}^{L^2,rel}$ | 2.46e-2 | 2.46e-2 | 9.05e-3 | 9.05e-3 | 3.98e-3 | 3.98e-3 |
| $e_{LOD-FEM}^{L^2,rel}$ | 2.46e-2 | | 9.05e-3 | | 3.98e-3 | |

Numerical Experiment 1: Stage 1 error in Stage 2 training



- ▶ **left plot:** $\eta_{a,\mu}$ detects dominant Stage 1 error and aborts enrichment.
- ▶ **right plot:** Ignoring Stage 1 error contributions leads to overfitted ROMs without further error decay.

Numerical Experiment 2

- ▶ $A_\mu = \sum_{q=1}^3 \mu_q A_q$
- ▶ $\mathcal{P} := [1, 5]^3$
- ▶ $|\mathcal{T}_h| = 67, 108, 864$
- ▶ $|\mathcal{T}_H| = 4, 096$
- ▶ $\kappa \approx 16$
- ▶ 1,024 processes
- ▶ $\varepsilon_1 = 0.01$ and $\varepsilon_2 = 0.02$.

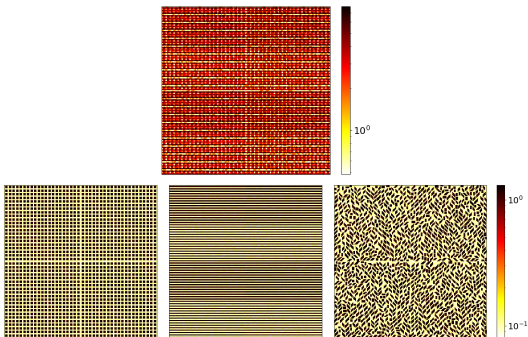


Figure: Coefficient A_μ on 4 coarse elements for $\mu = (1, 2, 3)^T$ (top center) and A_q for all $q = 1, \dots, 3$ (bottom).

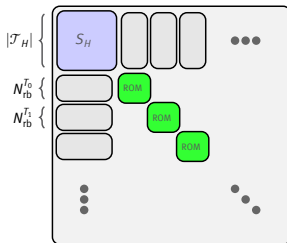
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| method | RBL0D | TSRBLOD |
|------------------------------------|---------------|---------------|
| $t_1^{\text{offline}}(T)$ | 10278 | 11289 |
| t_1^{offline} | 49436 | 54837 |
| t_2^{offline} | - | 9206 |
| t^{offline} | 49436 | 64043 |
| cum. size St.1 | 278528 | 193289 |
| av. size St.1 | 17.00 | 47.19 |
| size St.2 | - | 16 |
| storage | 409MB | 28KB |
| t^{LOD} (parallel) | 515 | |
| t^{online} (sequential) | 4.39 | 0.0005 |
| speed-up w.r.t LOD | 117 | 9.57e5 |
| $e_{\text{LOD}}^{H^1, \text{rel}}$ | 1.95e-5 | 4.43e-4 |
| $e_{\text{LOD}}^{L^2, \text{rel}}$ | 2.36e-5 | 4.49e-4 |

Some Remarks

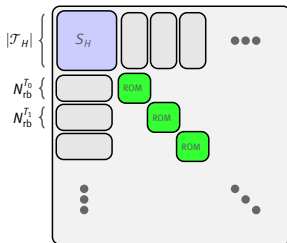
- ▶ \mathfrak{B}_μ has a sparse block structure and never needs to be assembled.
- ▶ Stage 1 can be performed in parallel without communication.
- ▶ Stage 2 completely $\dim V_h$ independent.



Block structure of \mathfrak{B}_μ w.r.t.
Stage 2 full-order space.

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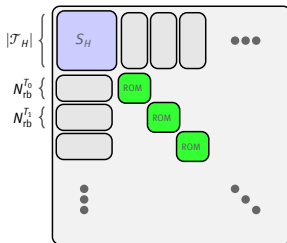
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- ▶ Adaptive strategies for locally decreasing ε_1 easily possible.



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- ▶ Stage 2 completely $\dim V_h$ independent.
- ▶ ε_1 can be chosen small to ensure stage 2 greedy succeeds (only offline time affected).
- ▶ Adaptive strategies for locally decreasing ε_1 easily possible.
- ▶ Error analysis also applies in the RBLOD case.
- ▶ Can be extended to other problem classes.



Block structure of \mathfrak{B}_μ w.r.t.
Stage 2 full-order space.

Thank you for your attention!

Keil, Rave, *An Online Efficient Two-Scale Reduced Basis Approach for the Localized Orthogonal Decomposition*, arXiv:2111.08643, 2021.