

An introduction to Hybrid High-Order methods

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- 1 Basics of HHO methods
- 2 Application to the incompressible Navier–Stokes problem

- 1 Basics of HHO methods
- 2 Application to the incompressible Navier–Stokes problem

- Capability of handling **general polyhedral meshes**
- Construction valid for **arbitrary space dimensions**
- Arbitrary **approximation order** (including $k = 0$)
- **Robustness** with respect to the variations of the physical coefficients
- Reduced **computational cost** after static condensation

$$N_{\text{dof},h} = \text{card}(\mathcal{F}_h^i) \binom{k+d-1}{d-1}$$

Polyhedral meshes

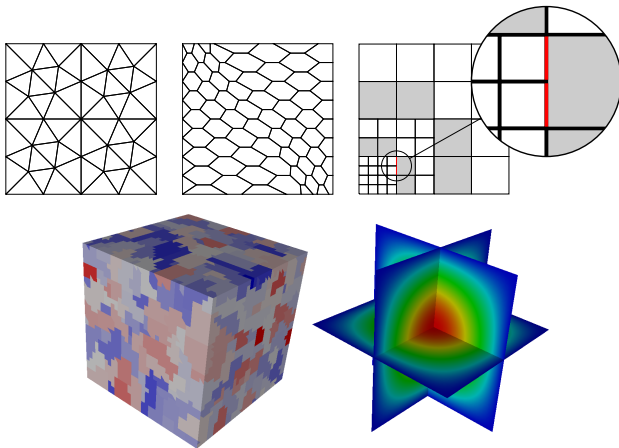


Figure: Admissible meshes in 2d and 3d, and HHO solution on the agglomerated 3d mesh

Model problem

- Let $\Omega \subset \mathbb{R}^d$, $d \geq 1$, denote a bounded, connected polyhedral domain
- For $f \in L^2(\Omega)$, we consider the **Poisson problem**

$$\begin{aligned} -\Delta u &= f && \text{in } \Omega \\ u &= 0 && \text{on } \partial\Omega \end{aligned}$$

- In weak form: Find $u \in H_0^1(\Omega)$ s.t.

$$a(u, v) := (\nabla u, \nabla v) = (f, v) \quad \forall v \in H_0^1(\Omega)$$

Projectors on local polynomial spaces

- At the core of HHO are **projectors on local polynomial spaces**
- With $X = T$ or $X = F$, the **L^2 -projector** $\pi_X^{0,l} : L^1(X) \rightarrow \mathbb{P}^l(X)$ is s.t.

$$(\pi_X^{0,l} v - v, w)_X = 0 \text{ for all } w \in \mathbb{P}^l(X)$$

- The **elliptic projector** $\pi_T^{1,l} : W^{1,1}(T) \rightarrow \mathbb{P}^l(T)$ is s.t.

$$(\nabla(\pi_T^{1,l} v - v), \nabla w)_T = 0 \text{ for all } w \in \mathbb{P}^l(T) \text{ and } (\pi_T^{1,l} v - v, 1)_T = 0$$

- Both $\pi_T^{0,l}$ and $\pi_T^{1,l}$ have **optimal approximation properties in $\mathbb{P}^l(T)$**
- See [DP and Droniou, 2017a, DP and Droniou, 2017b]

Computing $\pi_T^{1,k+1}$ from L^2 -projections of degree k

- The following integration by parts formula is valid for all $w \in C^\infty(\bar{T})$:

$$(\nabla v, \nabla w)_T = -(v, \Delta w)_T + \sum_{F \in \mathcal{F}_T} (v, \nabla w \cdot \mathbf{n}_{TF})_F$$

- Specializing it to $w \in \mathbb{P}^{k+1}(T)$, we can write

$$(\nabla \pi_T^{1,k+1} v, \nabla w)_T = -(\pi_T^{0,k} v, \Delta w)_T + \sum_{F \in \mathcal{F}_T} (\pi_F^{0,k} v|_F, \nabla w \cdot \mathbf{n}_{TF})_F$$

- Moreover, it can be easily seen that

$$(\pi_T^{1,k+1} v - v, 1)_T = (\pi_T^{1,k+1} v - \pi_T^{0,k} v, 1)_T = 0$$

- **Hence, $\pi_T^{1,k+1} v$ can be computed from $\pi_T^{0,k} v$ and $(\pi_F^{0,k} v|_F)_{F \in \mathcal{F}_T}$!**

Discrete unknowns

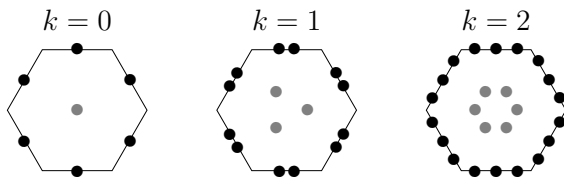


Figure: \underline{U}_T^k for $k \in \{0, 1, 2\}$

- Let a polynomial degree $k \geq 0$ be fixed
- For all $T \in \mathcal{T}_h$, we define the **local space of discrete unknowns**

$$\underline{U}_T^k := \{ \underline{v}_T = (v_T, (v_F)_{F \in \mathcal{F}_T}) : v_T \in \mathbb{P}^k(T) \text{ and } v_F \in \mathbb{P}^k(F) \quad \forall F \in \mathcal{F}_T \}$$

- The **local interpolator** $\underline{I}_T^k : H^1(T) \rightarrow \underline{U}_T^k$ is s.t., for all $v \in H^1(T)$,

$$\underline{I}_T^k v := (\pi_T^{0,k} v, (\pi_F^{0,k} v|_F)_{F \in \mathcal{F}_T})$$

Local potential reconstruction

- Let $T \in \mathcal{T}_h$. We define the local **potential reconstruction** operator

$$r_T^{k+1} : \underline{U}_T^k \rightarrow \mathbb{P}^{k+1}(T)$$

s.t. for all $\underline{v}_T \in \underline{U}_T^k$, $(r_T^{k+1} \underline{v}_T - v_T, 1)_T = 0$ and

$$(\nabla r_T^{k+1} \underline{v}_T, \nabla w)_T = -(v_T, \Delta w)_T + \sum_{F \in \mathcal{F}_T} (v_F, \nabla w \cdot \mathbf{n}_{TF})_F \quad \forall w \in \mathbb{P}^{k+1}(T)$$

- By construction, we have

$$r_T^{k+1} \circ \underline{I}_T^k = \pi_T^{1,k+1}$$

- $r_T^{k+1} \circ \underline{I}_T^k$ has therefore **optimal approximation properties** in $\mathbb{P}^{k+1}(T)$

- We would be tempted to approximate

$$a_T(u, v) \approx (\nabla r_T^{k+1} \underline{u}_T, \nabla r_T^{k+1} \underline{v}_T)_T$$

- This choice, however, is **not stable** in general. We consider instead

$$a_T(\underline{u}_T, \underline{v}_T) := (\nabla r_T^{k+1} \underline{u}_T, \nabla r_T^{k+1} \underline{v}_T)_T + s_T(\underline{u}_T, \underline{v}_T)$$

- The role of s_T is to ensure $\|\cdot\|_{1,T}$ -coercivity with

$$\|\underline{v}_T\|_{1,T}^2 := \|\nabla v_T\|_T^2 + \sum_{F \in \mathcal{F}_T} \frac{1}{h_F} \|v_F - v_T\|_F^2 \quad \forall \underline{v}_T \in \underline{U}_T^k$$

Assumption (Stabilization bilinear form)

The bilinear form $s_T : \underline{U}_T^k \times \underline{U}_T^k \rightarrow \mathbb{R}$ satisfies the following properties:

- **Symmetry and positivity.** s_T is symmetric and positive semidefinite.
- **Stability.** It holds, with hidden constant independent of h and T ,

$$a_T(\underline{v}_T, \underline{v}_T)^{\frac{1}{2}} \simeq \|\underline{v}_T\|_{1,T} \quad \forall \underline{v}_T \in \underline{U}_T^k.$$

- **Polynomial consistency.** For all $w \in \mathbb{P}^{k+1}(T)$ and all $\underline{v}_T \in \underline{U}_T^k$,

$$s_T(\underline{I}_T^k w, \underline{v}_T) = 0.$$

- The following stable choice **violates polynomial consistency**:

$$s_T^{\text{hdg}}(\underline{u}_T, \underline{v}_T) := \sum_{F \in \mathcal{F}_T} h_F^{-1} (u_F - u_T, v_F - v_T)_F$$

- To circumvent this problem, we penalize the **high-order differences** s.t.

$$(\delta_T^k \underline{v}_T, (\delta_{TF}^k \underline{v}_T)_{F \in \mathcal{F}_T}) := \underline{I}_T^k r_T^{k+1} \underline{v}_T - \underline{v}_T$$

- The classical HHO stabilization bilinear form reads

$$s_T(\underline{u}_T, \underline{v}_T) := \sum_{F \in \mathcal{F}_T} h_F^{-1} ((\delta_T^k - \delta_{TF}^k) \underline{u}_T, (\delta_T^k - \delta_{TF}^k) \underline{v}_T)_F$$

Discrete problem

- Define the **global space** with single-valued interface unknowns

$$\underline{U}_h^k := \left\{ \underline{v}_h = ((v_T)_{T \in \mathcal{T}_h}, (v_F)_{F \in \mathcal{F}_h}) : \right. \\ \left. v_T \in \mathbb{P}^k(T) \quad \forall T \in \mathcal{T}_h \text{ and } v_F \in \mathbb{P}^k(F) \quad \forall F \in \mathcal{F}_h \right\}$$

and its subspace with strongly enforced boundary conditions

$$\underline{U}_{h,0}^k := \{ \underline{v}_h \in \underline{U}_h^k : v_F \equiv 0 \quad \forall F \in \mathcal{F}_h^b \}$$

- The discrete problem reads: Find $\underline{u}_h \in \underline{U}_{h,0}^k$ s.t.

$$\mathbf{a}_h(\underline{u}_h, \underline{v}_h) := \sum_{T \in \mathcal{T}_h} \mathbf{a}_T(\underline{u}_T, \underline{v}_T) = \sum_{T \in \mathcal{T}_h} (f, v_T)_T \quad \forall \underline{v}_h \in \underline{U}_{h,0}^k$$

- Well-posedness** follows from coercivity and discrete Poincaré

Theorem (Energy-norm error estimate)

Assume $u \in H_0^1(\Omega) \cap H^{k+2}(\mathcal{T}_h)$. We have the following energy error estimate:

$$\|\nabla_h(r_h^{k+1}\underline{u}_h - u)\| + |\underline{u}_h|_{s,h} \lesssim h^{k+1} |u|_{H^{k+2}(\mathcal{T}_h)},$$

with $(r_h^{k+1}\underline{u}_h)|_T := r_T^{k+1}\underline{u}_T$ for all $T \in \mathcal{T}_h$ and $|\underline{u}_h|_{s,h}^2 := \sum_{T \in \mathcal{T}_h} s_T(\underline{u}_T, \underline{u}_T)$.

Theorem (Superclose L^2 -norm error estimate)

Further assuming *elliptic regularity* and $f \in H^1(\mathcal{T}_h)$ if $k = 0$,

$$\|r_h^{k+1}\underline{u}_h - u\| \lesssim h^{k+2} \mathcal{N}_k,$$

with $\mathcal{N}_0 := \|f\|_{H^1(\mathcal{T}_h)}$ and $\mathcal{N}_k := |u|_{H^{k+2}(\mathcal{T}_h)}$ for $k \geq 1$.

Static condensation I

- Fix a basis for $\underline{U}_{h,0}^k$ with functions supported by only one T or F
- Partition the discrete unknowns into element- and interface-based:

$$U_h = \begin{bmatrix} U_{\mathcal{T}_h} \\ U_{\mathcal{F}_h^i} \end{bmatrix}$$

- U_h solves the following linear system:

$$\begin{bmatrix} A_{\mathcal{T}_h \mathcal{T}_h} & A_{\mathcal{T}_h \mathcal{F}_h^i} \\ A_{\mathcal{F}_h^i \mathcal{T}_h} & A_{\mathcal{F}_h^i \mathcal{F}_h^i} \end{bmatrix} \begin{bmatrix} U_{\mathcal{T}_h} \\ U_{\mathcal{F}_h^i} \end{bmatrix} = \begin{bmatrix} F_{\mathcal{T}_h} \\ 0 \end{bmatrix}$$

- $A_{\mathcal{T}_h \mathcal{T}_h}$ is block-diagonal and SPD, hence inexpensive to invert

This remark suggests a two-step solution strategy:

- Element unknowns are eliminated solving the **local balances**

$$U_{\mathcal{T}_h} = A_{\mathcal{T}_h \mathcal{T}_h}^{-1} \left(F_{\mathcal{T}_h} - A_{\mathcal{T}_h \mathcal{F}_h^i} U_{\mathcal{F}_h^i} \right)$$

- Face unknowns are obtained solving the **global transmission problem**

$$A_h^{\text{SC}} U_{\mathcal{F}_h^i} = -A_{\mathcal{T}_h \mathcal{F}_h}^T A_{\mathcal{T}_h \mathcal{T}_h}^{-1} F_{\mathcal{T}_h}$$

with global system matrix

$$A_h^{\text{SC}} := A_{\mathcal{F}_h \mathcal{F}_h} - A_{\mathcal{T}_h \mathcal{F}_h}^T A_{\mathcal{T}_h \mathcal{T}_h}^{-1} A_{\mathcal{T}_h \mathcal{F}_h}$$

A_h^{SC} is SPD and its stencil involves neighbours through faces

Numerical examples

2d test case, smooth solution, uniform refinement

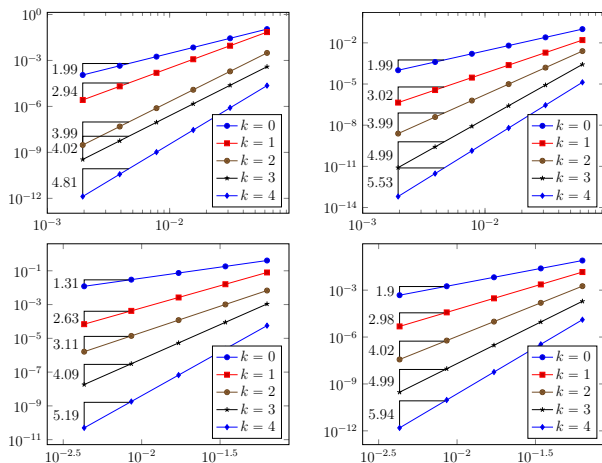


Figure: 2d test case, trigonometric solution. Energy (left) and L^2 -norm (right) of the error vs. h for uniformly refined **triangular** (top) and **hexagonal** (bottom) mesh families

Numerical examples I

3d industrial test case, adaptive refinement, cost assessment

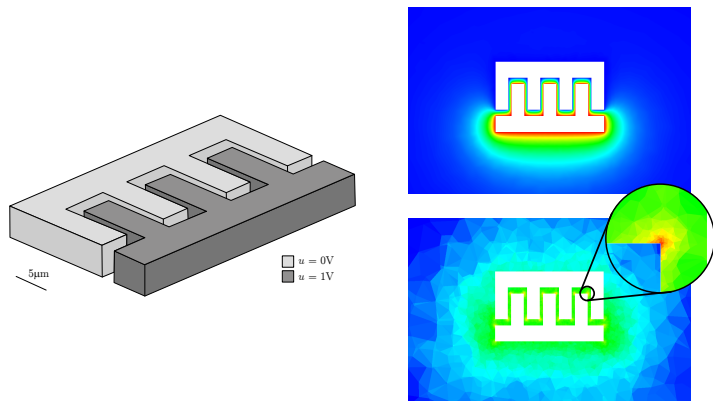
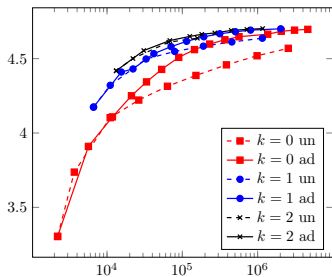


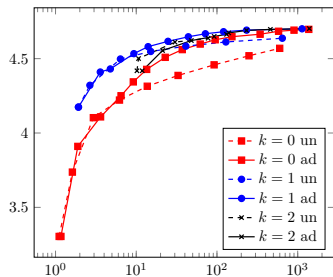
Figure: Geometry (left), numerical solution (right, top) and final adaptive mesh (right, bottom) for the comb-drive actuator test case [DP and Specogna, 2016]

Numerical examples II

3d industrial test case, adaptive refinement, cost assessment



(a) Capacitance vs. $N_{\text{dof},h}$

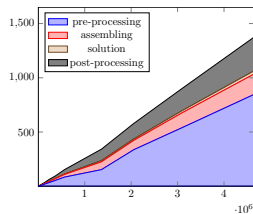


(b) Capacitance vs. computing time

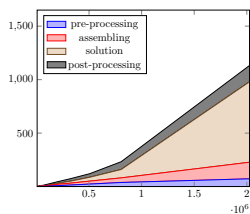
Figure: Results for the comb drive benchmark.

Numerical examples III

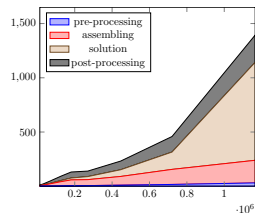
3d industrial test case, adaptive refinement, cost assessment



(a) $k = 0$



(b) $k = 1$



(c) $k = 2$

Figure: Computing wall time (s) vs. number of DOFs for the comb drive benchmark, AGMG solver.

Numerical examples I

3d test case, singular solution, adaptive coarsening

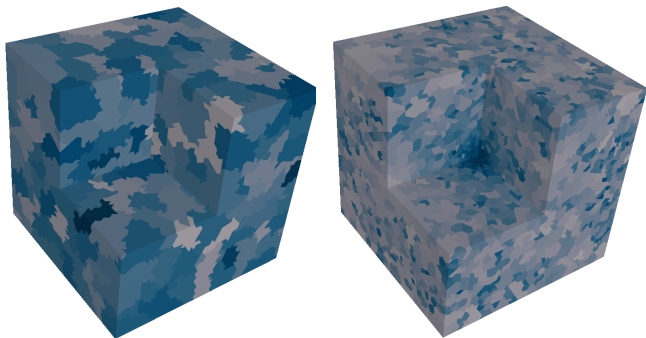
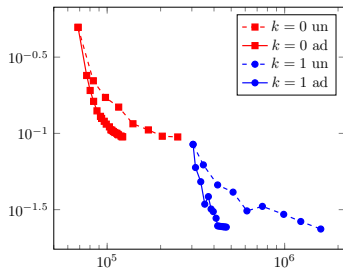


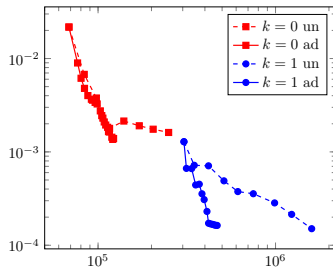
Figure: Fichera corner benchmark, adaptive mesh coarsening [DP and Specogna, 2016]

Numerical examples II

3d test case, singular solution, adaptive coarsening



(a) Energy-error vs. N_{dofs}



(b) L^2 -error vs. N_{dof}

Figure: Error vs. number of DOFs for the Fichera corner benchmark, adaptively coarsened meshes

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- Capability of handling **general polyhedral meshes**
- Construction valid for both $d = 2$ and $d = 3$
- Arbitrary **approximation order** (including $k = 0$)
- **Inf-sup stability** on general meshes
- **Robust handling** of dominant advection
- Reduced **computational cost** after static condensation

$$N_{\text{dof},h} = d \operatorname{card}(\mathcal{F}_h^i) \binom{k+d-1}{d-1} + \binom{k+d}{d}$$

A few references

- MHO for Stokes [Aghili, Boyaval, DP, 2015]
- Pressure-robust HHO for Stokes [DP, Ern, Linke, Schieweck, 2016]
- **HHO for Navier–Stokes [DP and Krell, 2017]**
- Péclet-robust HHO for Oseen [Aghili and DP, 2017]
- Darcy-robust HHO for Brinkman [Botti, DP, Droniou, 2018]

The incompressible Navier–Stokes equations I

- Let $d \in \{2, 3\}$, $\nu \in \mathbb{R}_+^*$, $\mathbf{f} \in L^2(\Omega)^d$, $\mathbf{U} := H_0^1(\Omega)^d$, and $P := L_0^2(\Omega)$
- The INS problem reads: Find $(\mathbf{u}, p) \in \mathbf{U} \times P$ s.t.

$$\begin{aligned} \nu a(\mathbf{u}, \mathbf{v}) + t(\mathbf{u}, \mathbf{u}, \mathbf{v}) + b(\mathbf{v}, p) &= \int_{\Omega} \mathbf{f} \cdot \mathbf{v} & \forall \mathbf{v} \in \mathbf{U}, \\ -b(\mathbf{u}, q) &= 0 & \forall q \in P, \end{aligned}$$

where

$$a(\mathbf{u}, \mathbf{v}) := \int_{\Omega} \nabla \mathbf{u} : \nabla \mathbf{v}, \quad b(\mathbf{v}, q) := - \int_{\Omega} (\nabla \cdot \mathbf{v}) q, \quad t(\mathbf{w}, \mathbf{u}, \mathbf{v}) := \int_{\Omega} \mathbf{v}^T \nabla \mathbf{u} \mathbf{w}$$

- Here, we have used the matrix-product notation, so that

$$\nabla \mathbf{u} \mathbf{u} = \sum_{j=1}^d u_j \partial_j \mathbf{u}$$

The incompressible Navier–Stokes equations II

- Integrating by parts and using $\mathbf{u} = \mathbf{0}$ on $\partial\Omega$ and $\nabla \cdot \mathbf{w} = 0$ in Ω , we get

$$t(\mathbf{w}, \mathbf{u}, \mathbf{v}) = \frac{1}{2} \int_{\Omega} \mathbf{v}^T \nabla \mathbf{u} \mathbf{w} - \frac{1}{2} \int_{\Omega} \mathbf{u}^T \nabla \mathbf{v} \mathbf{w}$$

- This shows that t is **non dissipative**: For all $\mathbf{w}, \mathbf{v} \in \mathbf{U}$ it holds

$$t(\mathbf{w}, \mathbf{v}, \mathbf{v}) = 0$$

Discrete spaces I

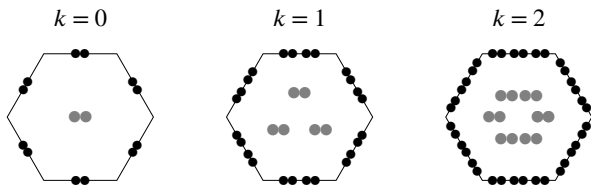


Figure: Local velocity space \underline{U}_T^k for $k \in \{0, 1, 2\}$

- For $k \geq 0$, we define the **global space of discrete unknowns**

$$\underline{U}_h^k := \left\{ \underline{v}_h = ((v_T)_{T \in \mathcal{T}_h}, (v_F)_{F \in \mathcal{F}_h}) : \right. \\ \left. v_T \in \mathbb{P}^k(T)^d \quad \forall T \in \mathcal{T}_h \text{ and } v_F \in \mathbb{P}^k(F)^d \quad \forall F \in \mathcal{F}_h \right\}$$

- The restriction to $T \in \mathcal{T}_h$ is denoted by \underline{U}_T^k , and $\underline{v}_T = (v_T, (v_F)_{F \in \mathcal{F}_T})$

- The **global interpolator** $\underline{\mathbf{I}}_h^k : H^1(\Omega)^d \rightarrow \underline{\mathbf{U}}_h^k$ is s.t. $\forall \mathbf{v} \in H^1(\Omega)^d$

$$\underline{\mathbf{I}}_h^k \mathbf{v} := ((\boldsymbol{\pi}_T^{0,k} \mathbf{v}|_T)_{T \in \mathcal{T}_h}, (\boldsymbol{\pi}_F^{0,k} \mathbf{v}|_F)_{F \in \mathcal{F}_h})$$

- The **velocity space** strongly accounting for boundary conditions is

$$\underline{\mathbf{U}}_{h,0}^k := \{ \underline{\mathbf{v}}_h \in \underline{\mathbf{U}}_h^k : \mathbf{v}_F = \mathbf{0} \quad \forall F \in \mathcal{F}_h^b \}$$

- The **discrete pressure space** is defined setting

$$\mathbf{P}_h^k := \left\{ q_h \in \mathbb{P}^k(\mathcal{T}_h) \mid \int_{\Omega} q_h = 0 \right\}$$

Reconstructions of differential operators

- For $l \geq 0$, the **gradient reconstruction** $\mathbf{G}_T^l : \underline{\mathbf{U}}_T^k \rightarrow \mathbb{P}^l(T)^{d \times d}$ is s.t.

$$\int_T \mathbf{G}_T^l \underline{\mathbf{v}}_T : \boldsymbol{\tau} = - \int_T \mathbf{v}_T \cdot (\nabla \cdot \boldsymbol{\tau}) + \sum_{F \in \mathcal{F}_T} \int_F \mathbf{v}_F \cdot (\boldsymbol{\tau} \mathbf{n}_{TF}) \quad \forall \boldsymbol{\tau} \in \mathbb{P}^l(T)^{d \times d}$$

- The **velocity reconstruction** $\mathbf{r}_T^{k+1} : \underline{\mathbf{U}}_T^k \rightarrow \mathbb{P}^{k+1}(T)^d$ is s.t.

$$\int_T (\nabla \mathbf{r}_T^{k+1} \underline{\mathbf{v}}_T - \mathbf{G}_T^k \underline{\mathbf{v}}_T) : \nabla \mathbf{w} = 0 \quad \forall \mathbf{w} \in \mathbb{P}^{k+1}(T)^d, \quad \int_T \mathbf{r}_T^{k+1} \underline{\mathbf{v}}_T - \mathbf{v}_T = \mathbf{0}$$

- Global reconstructions are defined setting for all $T \in \mathcal{T}_h$ and $\underline{\mathbf{v}}_h \in \underline{\mathbf{U}}_h^k$

$$(\mathbf{G}_h^l \underline{\mathbf{v}}_h)|_T := \mathbf{G}_T^l \underline{\mathbf{v}}_T, \quad (\mathbf{r}_h^{k+1} \underline{\mathbf{v}}_h)|_T := \mathbf{r}_T^{k+1} \underline{\mathbf{v}}_T, \quad \mathbf{D}_h^k \underline{\mathbf{v}}_h := \text{tr}(\mathbf{G}_h^k \underline{\mathbf{v}}_h)$$

- The **viscous term** is discretized by means of the bilinear form a_h s.t.

$$a_h(\underline{\mathbf{u}}_h, \underline{\mathbf{v}}_h) := \int_{\Omega} \mathbf{G}_h^k \underline{\mathbf{u}}_h : \mathbf{G}_h^k \underline{\mathbf{v}}_h + s_h(\underline{\mathbf{u}}_h, \underline{\mathbf{v}}_h)$$

- As in the scalar case, several possible choices for s_h ensure that

$$C_a^{-1} \|\underline{\mathbf{v}}_h\|_{1,h}^2 \leq a_h(\underline{\mathbf{v}}_h, \underline{\mathbf{v}}_h) \leq C_a \|\underline{\mathbf{v}}_h\|_{1,h}^2 \quad \forall \underline{\mathbf{v}}_h \in \underline{\mathbf{U}}_h^k$$

with real number C_a independent of h and of the problem data

- **Variable viscosity** can be treated following [DP and Ern, 2015]

Pressure-velocity coupling

- The pressure-velocity coupling is realized by means of the bilinear

$$b_h(\underline{\mathbf{v}}_h, q_h) := - \int_{\Omega} D_h^k \underline{\mathbf{v}}_h q_h$$

- A crucial point is that b_h satisfies the following **inf-sup condition**

$$\forall q_h \in P_h^k, \quad \beta \|q_h\|_{L^2(\Omega)} \leq \sup_{\underline{\mathbf{v}}_h \in \underline{U}_{h,0}^k, \|\underline{\mathbf{v}}_h\|_{1,h}=1} b_h(\underline{\mathbf{v}}_h, q_h)$$

- This stability result is valid on **general meshes** and for any $k \geq 0$

Convective term I

- Recall the skew-symmetric expression of t :

$$t(\mathbf{w}, \mathbf{u}, \mathbf{v}) = \frac{1}{2} \int_{\Omega} \mathbf{v}^T \nabla \mathbf{u} \mathbf{w} - \frac{1}{2} \int_{\Omega} \mathbf{u}^T \nabla \mathbf{v} \mathbf{w}$$

- Inspired by this reformulation, we set

$$t_h(\underline{\mathbf{w}}_h, \underline{\mathbf{u}}_h, \underline{\mathbf{v}}_h) := \frac{1}{2} \int_{\Omega} \mathbf{v}_h^T \mathbf{G}_h^{2k} \underline{\mathbf{u}}_h \mathbf{w}_h - \frac{1}{2} \int_{\Omega} \mathbf{u}_h^T \mathbf{G}_h^{2k} \underline{\mathbf{v}}_h \mathbf{w}_h$$

- By design, t_h is **non dissipative**: For all $\underline{\mathbf{w}}_h, \underline{\mathbf{v}}_h$,

$$t_h(\underline{\mathbf{w}}_h, \underline{\mathbf{v}}_h, \underline{\mathbf{v}}_h) = 0$$

Convective term II

- In practice, one **does not need to actually compute** \mathbf{G}_h^{2k}
- In fact, expanding \mathbf{G}_h^{2k} according to its definition, we have

$$t_h(\underline{\mathbf{w}}_h, \underline{\mathbf{u}}_h, \underline{\mathbf{v}}_h) = \sum_{T \in \mathcal{T}_h} t_T(\underline{\mathbf{w}}_T, \underline{\mathbf{u}}_T, \underline{\mathbf{v}}_T),$$

where, for all $T \in \mathcal{T}_h$,

$$\begin{aligned} t_T(\underline{\mathbf{w}}_T, \underline{\mathbf{u}}_T, \underline{\mathbf{v}}_T) &:= -\frac{1}{2} \int_T \mathbf{u}_T^T \nabla \mathbf{v}_T \mathbf{w}_T + \frac{1}{2} \sum_{F \in \mathcal{F}_T} \int_F (\mathbf{u}_F \cdot \mathbf{v}_T) (\mathbf{w}_T \cdot \mathbf{n}_{TF}) \\ &+ \frac{1}{2} \int_T \mathbf{v}_T^T \nabla \mathbf{u}_T \mathbf{w}_T - \frac{1}{2} \sum_{F \in \mathcal{F}_T} \int_F (\mathbf{v}_F \cdot \mathbf{u}_T) (\mathbf{w}_T \cdot \mathbf{n}_{TF}) \end{aligned}$$

Discrete problem I

The discrete problem reads: Find $(\underline{\mathbf{u}}_h, p_h) \in \underline{\mathbf{U}}_{h,0}^k \times P_h^k$ s.t.

$$\begin{aligned} \mathbf{va}_h(\underline{\mathbf{u}}_h, \underline{\mathbf{v}}_h) + \mathbf{t}_h(\underline{\mathbf{u}}_h, \underline{\mathbf{u}}_h, \underline{\mathbf{v}}_h) + \mathbf{b}_h(\underline{\mathbf{v}}_h, p_h) &= \int_{\Omega} \mathbf{f} \cdot \underline{\mathbf{v}}_h & \forall \underline{\mathbf{v}}_h \in \underline{\mathbf{U}}_{h,0}^k, \\ -\mathbf{b}_h(\underline{\mathbf{u}}_h, q_h) &= 0 & \forall q_h \in P_h^k \end{aligned}$$

Theorem (Existence and a priori bounds)

There exists a solution $(\underline{\mathbf{u}}_h, p_h) \in \underline{\mathbf{U}}_{h,0}^k \times P_h^k$ such that

$$\begin{aligned}\|\underline{\mathbf{u}}_h\|_{1,h} &\leq C_a C_s \nu^{-1} \|\mathbf{f}\|, \\ \|p_h\| &\leq C (\|\mathbf{f}\| + \nu^{-2} \|\mathbf{f}\|^2),\end{aligned}$$

with C_s discrete Poincaré constant, and $C > 0$ independent of h and ν .

Theorem (Uniqueness of the discrete solution)

Assume that the forcing term verifies

$$\|\mathbf{f}\| \leq \frac{\nu^2}{2C_a^2 C_t C_s}$$

with C_t continuity constant of \mathfrak{t}_h . Then, the solution is unique.

Theorem (Convergence to minimal regularity solutions)

It holds up to a subsequence, as $h \rightarrow 0$,

- $\mathbf{u}_h \rightarrow \mathbf{u}$ strongly in $L^p(\Omega)^d$ for $p \in [1, +\infty)$ if $d = 2$, $p \in [1, 6)$ if $d = 3$;
- $\mathbf{G}_h^k \underline{\mathbf{u}}_h \rightarrow \nabla \mathbf{u}$ strongly in $L^2(\Omega)^{d \times d}$;
- $s_h(\underline{\mathbf{u}}_h, \underline{\mathbf{u}}_h) \rightarrow 0$;
- $p_h \rightarrow p$ strongly in $L^2(\Omega)$.

If the exact solution is unique, the whole sequence converges.

Key tools: discrete **Sobolev embeddings** and **Rellick–Kondrachov compactness results** from [DP and Droniou, 2017a]

Convergence II

Theorem (Convergence rates for small data)

Assume uniqueness for both $(\underline{\mathbf{u}}_h, p_h)$ and (\mathbf{u}, p) . Assume, moreover, the additional regularity $(\mathbf{u}, p) \in H^{k+2}(\Omega)^d \times H^{k+1}(\Omega)$, as well as

$$\|\mathbf{f}\| \leq \frac{\nu^2}{2C_I C_a C_t (1 + C_P^2)},$$

with C_a and C_t as above, C_I boundedness constant of $\underline{\mathbf{I}}_h^k$, and C_P continuous Poincaré constant. Then, with hidden constant independent of both h and ν ,

$$\|\underline{\mathbf{u}}_h - \underline{\mathbf{I}}_h^k \mathbf{u}\|_{1,h} + \nu^{-1} \|p_h - \pi_h^{0,k} p\|_{L^2(\Omega)} \lesssim h^{k+1} \mathcal{N}_{\mathbf{u},p}.$$

with $\mathcal{N}_{\mathbf{u},p} := (1 + \nu^{-1} \|\mathbf{u}\|_{H^2(\Omega)^d}) \|\mathbf{u}\|_{H^{k+2}(\Omega)^d} + \nu^{-1} \|p\|_{H^{k+1}(\Omega)}$,

Static condensation

- Partition the discrete velocity unknowns as before, and the pressure unknowns into **average value + oscillations** inside each element
- At each iteration, the linear system has the form

$$\begin{bmatrix} A_{\mathcal{T}_h \mathcal{T}_h} & \tilde{B}_{\mathcal{T}_h} & A_{\mathcal{T}_h \mathcal{F}_h^i} & \bar{B}_{\mathcal{T}_h} \\ A_{\mathcal{F}_h^i \mathcal{T}_h} & \tilde{B}_{\mathcal{F}_h^i} & A_{\mathcal{F}_h^i \mathcal{F}_h^i} & \bar{B}_{\mathcal{F}_h^i} \\ \tilde{B}_{\mathcal{T}_h}^T & 0 & \tilde{B}_{\mathcal{F}_h^i}^T & 0 \\ \bar{B}_{\mathcal{T}_h}^T & 0 & \bar{B}_{\mathcal{F}_h^i}^T & 0 \end{bmatrix} \begin{bmatrix} U_{\mathcal{T}_h} \\ \tilde{P}_{\mathcal{T}_h} \\ U_{\mathcal{F}_h^i} \\ \bar{P}_{\mathcal{T}_h} \end{bmatrix} = \begin{bmatrix} F_{\mathcal{T}_h} \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

- Static condensation of $U_{\mathcal{T}_h}$ and $\tilde{P}_{\mathcal{T}_h}$ is possible
- **Impact of static condensation on the global matrix?**

Numerical example: Kovaszny flow

- We consider the exact solution of [Kovaszny, 1948].
- Let $\Omega := (-0.5, 1.5) \times (0, 2)$ and set

$$\text{Re} := (2\nu)^{-1}, \quad \lambda := \text{Re} - (\text{Re}^2 + 4\pi^2)^{\frac{1}{2}}$$

- The components of the velocity are given by

$$u_1(\mathbf{x}) := 1 - \exp(\lambda x_1) \cos(2\pi x_2), \quad u_2(\mathbf{x}) := \frac{\lambda}{2\pi} \exp(\lambda x_1) \sin(2\pi x_2),$$

and pressure given by

$$p(\mathbf{x}) := -\frac{1}{2} \exp(2\lambda x_1) + \frac{\lambda}{2} (\exp(4\lambda) - 1)$$

Numerical example: Kovasznay flow

Cartesian mesh family, $\nu = 0.1$

h	$\ \underline{u}_h - \hat{\underline{u}}_h\ _{1,h}$	OCR	$\ \mathbf{u}_h - \hat{\mathbf{u}}_h\ $	OCR	$\ p_h - \hat{p}_h\ $	OCR
$k = 0$						
0.13	1.02	—	0.33	—	1.84	—
$6.25 \cdot 10^{-2}$	0.55	0.89	0.17	0.99	0.21	3.14
$3.12 \cdot 10^{-2}$	0.34	0.68	$4.31 \cdot 10^{-2}$	1.94	$4.36 \cdot 10^{-2}$	2.26
$1.56 \cdot 10^{-2}$	0.19	0.86	$1.09 \cdot 10^{-2}$	1.98	$1.37 \cdot 10^{-2}$	1.67
$7.81 \cdot 10^{-3}$	$9.72 \cdot 10^{-2}$	0.96	$2.7 \cdot 10^{-3}$	2.02	$3.78 \cdot 10^{-3}$	1.86
$k = 1$						
0.13	0.45	—	0.15	—	0.44	—
$6.25 \cdot 10^{-2}$	0.24	0.94	$3.39 \cdot 10^{-2}$	2.11	$3.45 \cdot 10^{-2}$	3.68
$3.12 \cdot 10^{-2}$	$6.46 \cdot 10^{-2}$	1.86	$4.26 \cdot 10^{-3}$	2.99	$8.58 \cdot 10^{-3}$	2
$1.56 \cdot 10^{-2}$	$1.78 \cdot 10^{-2}$	1.86	$5.58 \cdot 10^{-4}$	2.93	$1.23 \cdot 10^{-3}$	2.8
$7.81 \cdot 10^{-3}$	$4.65 \cdot 10^{-3}$	1.94	$7.11 \cdot 10^{-5}$	2.98	$1.87 \cdot 10^{-4}$	2.72
$k = 2$						
0.13	0.25	—	$6.41 \cdot 10^{-2}$	—	$9.8 \cdot 10^{-2}$	—
$6.25 \cdot 10^{-2}$	$4.83 \cdot 10^{-2}$	2.34	$5.81 \cdot 10^{-3}$	3.46	$7.55 \cdot 10^{-3}$	3.7
$3.12 \cdot 10^{-2}$	$7.11 \cdot 10^{-3}$	2.76	$3.45 \cdot 10^{-4}$	4.06	$7.71 \cdot 10^{-4}$	3.28
$1.56 \cdot 10^{-2}$	$1.01 \cdot 10^{-3}$	2.82	$2.07 \cdot 10^{-5}$	4.06	$7 \cdot 10^{-5}$	3.46
$7.81 \cdot 10^{-3}$	$1.34 \cdot 10^{-4}$	2.92	$1.25 \cdot 10^{-6}$	4.06	$6.54 \cdot 10^{-6}$	3.43
$k = 3$						
0.13	$7.84 \cdot 10^{-2}$	—	$2.1 \cdot 10^{-2}$	—	$3.46 \cdot 10^{-2}$	—
$6.25 \cdot 10^{-2}$	$7.5 \cdot 10^{-3}$	3.39	$8.03 \cdot 10^{-4}$	4.71	$1.39 \cdot 10^{-3}$	4.64
$3.12 \cdot 10^{-2}$	$5.11 \cdot 10^{-4}$	3.87	$2.52 \cdot 10^{-5}$	4.98	$7.31 \cdot 10^{-5}$	4.24
$1.56 \cdot 10^{-2}$	$3.43 \cdot 10^{-5}$	3.9	$8.15 \cdot 10^{-7}$	4.95	$3.87 \cdot 10^{-6}$	4.24
$7.81 \cdot 10^{-3}$	$2.22 \cdot 10^{-6}$	3.96	$2.59 \cdot 10^{-8}$	4.98	$2.17 \cdot 10^{-7}$	4.16

Numerical example: Kovasznay flow

Hexagonal mesh family, $\nu = 0.1$

h	$\ \underline{u}_h - \hat{\underline{u}}_h\ _{1,h}$	OCR	$\ \mathbf{u}_h - \hat{\mathbf{u}}_h\ $	OCR	$\ p_h - \hat{p}_h\ $	OCR
$k = 0$						
0.14	1.64	—	0.62	—	2.1	—
$7.33 \cdot 10^{-2}$	0.64	1.44	0.19	1.81	0.24	3.31
$3.69 \cdot 10^{-2}$	0.44	0.56	$7.12 \cdot 10^{-2}$	1.42	$9.99 \cdot 10^{-2}$	1.28
$1.85 \cdot 10^{-2}$	0.25	0.79	$2.32 \cdot 10^{-2}$	1.62	$3.94 \cdot 10^{-2}$	1.35
$9.27 \cdot 10^{-3}$	0.13	0.91	$6.7 \cdot 10^{-3}$	1.8	$1.32 \cdot 10^{-2}$	1.58
$k = 1$						
0.14	0.53	—	0.22	—	0.28	—
$7.33 \cdot 10^{-2}$	0.22	1.32	$3.95 \cdot 10^{-2}$	2.64	$5.25 \cdot 10^{-2}$	2.58
$3.69 \cdot 10^{-2}$	$7.26 \cdot 10^{-2}$	1.63	$4.81 \cdot 10^{-3}$	3.07	$1.26 \cdot 10^{-2}$	2.08
$1.85 \cdot 10^{-2}$	$1.96 \cdot 10^{-2}$	1.9	$5.81 \cdot 10^{-4}$	3.06	$2.37 \cdot 10^{-3}$	2.42
$9.27 \cdot 10^{-3}$	$5.07 \cdot 10^{-3}$	1.96	$6.75 \cdot 10^{-5}$	3.12	$4.07 \cdot 10^{-4}$	2.55
$k = 2$						
0.14	0.28	—	$7.84 \cdot 10^{-2}$	—	0.11	—
$7.33 \cdot 10^{-2}$	$5.23 \cdot 10^{-2}$	2.56	$6.37 \cdot 10^{-3}$	3.84	$1.19 \cdot 10^{-2}$	3.39
$3.69 \cdot 10^{-2}$	$8.32 \cdot 10^{-3}$	2.68	$5.32 \cdot 10^{-4}$	3.62	$1.7 \cdot 10^{-3}$	2.84
$1.85 \cdot 10^{-2}$	$1.16 \cdot 10^{-3}$	2.85	$3.74 \cdot 10^{-5}$	3.85	$2.04 \cdot 10^{-4}$	3.07
$9.27 \cdot 10^{-3}$	$1.52 \cdot 10^{-4}$	2.94	$2.44 \cdot 10^{-6}$	3.95	$2.61 \cdot 10^{-5}$	2.98
$k = 3$						
0.14	$7.1 \cdot 10^{-2}$	—	$1.56 \cdot 10^{-2}$	—	$2.23 \cdot 10^{-2}$	—
$7.33 \cdot 10^{-2}$	$9.66 \cdot 10^{-3}$	3.05	$1.1 \cdot 10^{-3}$	4.05	$2.31 \cdot 10^{-3}$	3.47
$3.69 \cdot 10^{-2}$	$8.97 \cdot 10^{-4}$	3.46	$5.36 \cdot 10^{-5}$	4.4	$1.7 \cdot 10^{-4}$	3.8
$1.85 \cdot 10^{-2}$	$6.8 \cdot 10^{-5}$	3.74	$2.13 \cdot 10^{-6}$	4.67	$1.08 \cdot 10^{-5}$	3.99
$9.27 \cdot 10^{-3}$	$4.68 \cdot 10^{-6}$	3.87	$7.6 \cdot 10^{-8}$	4.82	$6.69 \cdot 10^{-7}$	4.03

Numerical example: Kovasznay flow

Hexagonal mesh family, HDG trilinear form, $\nu = 0.1$

h	$\ \underline{\mathbf{u}}_h - \hat{\underline{\mathbf{u}}}_h\ _{1,h}$	OCR	$\ \mathbf{u}_h - \hat{\mathbf{u}}_h\ $	OCR	$\ p_h - \hat{p}_h\ $	OCR
$k = 1$						
0.14			Not converged			
$7.33 \cdot 10^{-2}$	0.22	—	$3.99 \cdot 10^{-2}$	—	$4.83 \cdot 10^{-2}$	—
$3.69 \cdot 10^{-2}$	$7.01 \cdot 10^{-2}$	1.65	$4.94 \cdot 10^{-3}$	3.04	$9.91 \cdot 10^{-3}$	2.31
$1.85 \cdot 10^{-2}$	$1.94 \cdot 10^{-2}$	1.86	$5.87 \cdot 10^{-4}$	3.09	$1.94 \cdot 10^{-3}$	2.36
$9.27 \cdot 10^{-3}$	$5.04 \cdot 10^{-3}$	1.95	$6.64 \cdot 10^{-5}$	3.15	$3.5 \cdot 10^{-4}$	2.48
$k = 2$						
0.14			Not converged			
$7.33 \cdot 10^{-2}$	$4.96 \cdot 10^{-2}$	—	$6.36 \cdot 10^{-3}$	—	$9.52 \cdot 10^{-3}$	—
$3.69 \cdot 10^{-2}$	$8.38 \cdot 10^{-3}$	2.59	$5.52 \cdot 10^{-4}$	3.56	$1.38 \cdot 10^{-3}$	2.81
$1.85 \cdot 10^{-2}$	$1.18 \cdot 10^{-3}$	2.84	$3.92 \cdot 10^{-5}$	3.83	$1.73 \cdot 10^{-4}$	3.01
$9.27 \cdot 10^{-3}$	$1.55 \cdot 10^{-4}$	2.94	$2.58 \cdot 10^{-6}$	3.94	$2.25 \cdot 10^{-5}$	2.95
$k = 3$						
0.14	$6.69 \cdot 10^{-2}$	—	$1.52 \cdot 10^{-2}$	—	$1.65 \cdot 10^{-2}$	—
$7.33 \cdot 10^{-2}$	$9.61 \cdot 10^{-3}$	2.97	$1.1 \cdot 10^{-3}$	4.01	$1.91 \cdot 10^{-3}$	3.3
$3.69 \cdot 10^{-2}$	$9.14 \cdot 10^{-4}$	3.43	$5.56 \cdot 10^{-5}$	4.35	$1.5 \cdot 10^{-4}$	3.71
$1.85 \cdot 10^{-2}$	$6.99 \cdot 10^{-5}$	3.72	$2.24 \cdot 10^{-6}$	4.65	$9.86 \cdot 10^{-6}$	3.94
$9.27 \cdot 10^{-3}$	$4.83 \cdot 10^{-6}$	3.87	$8.01 \cdot 10^{-8}$	4.82	$6.17 \cdot 10^{-7}$	4.01

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