## Hybrid high-order methods

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#### ENPC and INRIA, Paris, France joint work with E. Burman (UCL), G. Delay (Sorbonne), O. Duran (Bergen) collaboration and support: CEA

#### MFET, Mülheim an der Ruhr, 22 August 2023

## Outline

Hybrid high-order (HHO) methods ...

- In a nutshell
- Links to other methods
- Wave propagation problems

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- Links to other methods
- Wave propagation problems
- Seminal references: [Di Pietro, AE, Lemaire 14; Di Pietro, AE 15]
- Two textbooks
  - HHO on polytopal meshes [Di Pietro, Droniou 20]
  - A primer with application to solid mechanics [Cicuttin, AE, Pignet 21]



# HHO in a nutshell

## Basic ideas

- Degrees of freedom (dofs) located on mesh cells and faces
- Let us start with polynomials of the same degree *k* ≥ 0 on cells and faces



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- The global problem is assembled cellwise as in FEM
- Generalization to higher order of ideas from Hybrid FV and Hybrid Mimetic Mixed methods [Eymard, Gallouet, Herbin 10; Droniou et al. 10]

#### Gradient reconstruction and stabilization

• Mesh cell  $T \in \mathcal{T}$ , cell dofs  $u_T \in \mathbb{P}^k(T)$ , face dofs  $u_{\partial T} \in \mathbb{P}^k(\mathcal{F}_{\partial T})$ 

$$\hat{u}_T = (u_T, u_{\partial T}) \in \hat{U}_T := \mathbb{P}^k(T) \times \mathbb{P}^k(\mathcal{F}_{\partial T})$$

• Local potential reconstruction  $R_T : \hat{U}_T \to \mathbb{P}^{k+1}(T)$  s.t.

 $(\nabla R_T(\hat{u}_T), \nabla q)_T = -(u_T, \Delta q)_T + (u_{\partial T}, \nabla q \cdot \mathbf{n}_T)_{\partial T}, \quad \forall q \in \mathbb{P}^{k+1}(T)/\mathbb{R}$ together with  $(R_T(\hat{u}_T), 1)_T = (u_T, 1)_T$ 

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- Local gradient reconstruction  $\mathbf{G}_T(\hat{u}_T) := \nabla R_T(\hat{u}_T) \in \nabla \mathbb{P}^{k+1}(T)$
- Local stabilization operator acting on  $\delta_{\hat{u}_T} := u_T |_{\partial T} u_{\partial T}$

$$S_{\partial T}(\delta_{\hat{u}_T}) := \prod_{\partial T}^k \left( \delta_{\hat{u}_T} - \underbrace{\left( (I - \prod_T^k) R_T(0, \delta_{\hat{u}_T}) \right) |_{\partial T}}_{\mathcal{O}} \right)$$

high-order correction

Taking  $S_{\partial T}(\delta_{\hat{u}_T}) := \delta_{\hat{u}_T}$  is suboptimal ...

• Local bilinear form for Poisson model problem

 $a_T(\hat{u}_T, \hat{w}_T) := (\mathbf{G}_T(\hat{u}_T), \mathbf{G}_T(\hat{w}_T))_T + h_T^{-1}(S_{\partial T}(\delta_{\hat{u}_T}), S_{\partial T}(\delta_{\hat{w}_T}))_{\partial T}$ 

(recall  $\delta_{\hat{u}_T} := u_T |_{\partial T} - u_{\partial T}$ )

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• Stability and boundedness

$$\alpha \|\hat{u}_{T}\|_{\hat{U}_{T}}^{2} \leq a_{T}(\hat{u}_{T}, \hat{u}_{T}) \leq \omega \|\hat{u}_{T}\|_{\hat{U}_{T}}^{2}, \quad \forall \hat{u}_{T} \in \hat{U}_{T}$$
  
with  $\|\hat{u}_{T}\|_{\hat{U}_{T}}^{2} := \|\nabla u_{T}\|_{T}^{2} + h_{T}^{-1} \|\delta_{\hat{u}_{T}}\|_{\partial T}^{2}$ 

#### Assembly of discrete problem

• Global dofs  $\hat{u}_h = (u_{\mathcal{T}}, u_{\mathcal{F}}) \ (\mathcal{T} := \{\text{mesh cells}\}, \mathcal{F} := \{\text{mesh faces}\})$ 

$$\hat{U}_h := \mathbb{P}^k(\mathcal{T}) \times \mathbb{P}^k(\mathcal{F}), \quad \mathbb{P}^k(\mathcal{T}) := \sum_{T \in \mathcal{T}} \mathbb{P}^k(T), \quad \mathbb{P}^k(\mathcal{F}) := \sum_{F \in \mathcal{F}} \mathbb{P}^k(F)$$

• Dirichlet conditions enforced on face boundary dofs

$$\hat{U}_{h0} := \{ \hat{v}_h \in \hat{U}_h \mid v_F = 0 \; \forall F \subset \partial \Omega \}$$

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• Dirichlet conditions enforced on face boundary dofs

$$\hat{U}_{h0} := \{ \hat{v}_h \in \hat{U}_h \mid v_F = 0 \; \forall F \subset \partial \Omega \}$$

• Discrete problem: Find  $\hat{u}_h \in \hat{U}_{h0}$  s.t.

$$a_h(\hat{u}_h, \hat{w}_h) := \sum_{T \in \mathcal{T}} a_T(\hat{u}_T, \hat{w}_T) = (f, w_{\mathcal{T}})_{\Omega}, \quad \forall \hat{w}_h \in \hat{U}_{h0}$$

(only cell component of test function used on rhs)

#### Algebraic realization and static condensation

• Algebraic realization

$$\begin{bmatrix} \mathsf{A}_{\mathcal{T}\mathcal{T}} & \mathsf{A}_{\mathcal{T}\mathcal{F}} \\ \mathsf{A}_{\mathcal{F}\mathcal{T}} & \mathsf{A}_{\mathcal{F}\mathcal{F}} \end{bmatrix} \begin{bmatrix} \mathsf{U}_{\mathcal{T}} \\ \mathsf{U}_{\mathcal{F}} \end{bmatrix} = \begin{bmatrix} \mathsf{F}_{\mathcal{T}} \\ 0 \end{bmatrix}$$

 $\implies$  submatrix  $A_{\mathcal{TT}}$  is block-diagonal!

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- Cell dofs can be eliminated locally by static condensation
  - global problem couples only face dofs
  - cell dofs recovered by local post-processing
- Summary



#### Main assets of HHO methods

• General meshes: polytopal cells, hanging nodes



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#### Local conservation

- optimally convergent and algebraically balanced fluxes on faces
- as any face-based method, balance at the cell level

#### • Attractive computational costs

- only face dofs are globally coupled
- compact stencil (slightly less compact than DG though)

#### Error estimates

- Smooth solutions (in  $H^{k+2}(\Omega)$ )
  - $O(h^{k+1}) H^1$ -error estimate (face dofs of order  $k \ge 0$ )
  - $O(h^{k+2})$  L<sup>2</sup>-error estimate (with full elliptic regularity)

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  - $O(h^{k+2})$  L<sup>2</sup>-error estimate (with full elliptic regularity)
- Less regularity
  - $O(h^t) H^1$ -error estimate if  $u \in H^{1+t}(\Omega), t \in (\frac{1}{2}, k+1]$
  - for  $t \in (0, \frac{1}{2})$ , see [AE, Guermond 21 (FoCM)]
  - for  $f \in H^{-1}(\Omega)$ , see [AE, Zanotti 20 (IMAJNA)]

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  - for  $t \in (0, \frac{1}{2})$ , see [AE, Guermond 21 (FoCM)]
  - for  $f \in H^{-1}(\Omega)$ , see [AE, Zanotti 20 (IMAJNA)]
- Main consistency property: Introduce reduction operator

$$\hat{l}_T : H^1(T) \to \hat{U}_T, \qquad \hat{l}_T(v) := (\Pi^k_T(v), \Pi^k_{\partial T}(v|_{\partial T}))$$

Then we have

• 
$$h_T^{-1} \| v - R_T(\hat{l}_T(v)) \|_T + \| \nabla (v - R_T(\hat{l}_T(v))) \|_T \leq h_T^{k+1} |v|_{H^{k+2}(T)}$$
  
•  $h_T^{-\frac{1}{2}} \| S_{\partial T}(\hat{l}_T(v)) \|_{\partial T} \leq h_T^{k+1} |v|_{H^{k+2}(T)}$ 

#### Variants

• Variant on gradient reconstruction  $\mathbf{G}_T : \hat{U}_T \to \mathbb{P}^k(T; \mathbb{R}^d)$  s.t.

$$(\mathbf{G}_T(\hat{u}_T), \mathbf{q})_T = -(\mathbf{u}_T, \operatorname{div} \mathbf{q})_T + (\mathbf{u}_{\partial T}, \mathbf{q} \cdot \mathbf{n}_T)_{\partial T}, \quad \forall \mathbf{q} \in \mathbb{P}^k(T; \mathbb{R}^d)$$

- same scalar mass matrix for each component of  $\mathbf{G}_T(\hat{u}_T)$
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   [Di Pietro, Droniou 17; Botti, Di Pietro, Sochala 17; Abbas, AE, Pignet 18]
- Variants on cell dofs and stabilization
  - mixed-order setting:  $k \ge 0$  for face dofs and (k + 1) for cell dofs
  - this variant allows for the simpler Lehrenfeld-Schöberl HDG stabilization

$$S_{\partial T}(\delta_{\hat{u}_T}) := \Pi^k_{\partial T}(\delta_{\hat{u}_T})$$

• another variant is  $k \ge 1$  for face dofs and (k - 1) for cell dofs

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- One idea is to use unfitted meshes
  - curved interface can cut arbitrarily through mesh cells
  - numerical method must deal with ill cut cells

## HHO on unfitted meshes

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- HHO works optimally on cells with planar faces
- One idea is to use unfitted meshes
  - curved interface can cut arbitrarily through mesh cells
  - numerical method must deal with ill cut cells
- Well developed paradigm for unfitted FEM
  - double unknowns in cut cells and use a consistent Nitsche's penalty technique to enforce jump conditions [Hansbo, Hansbo 02]
  - ghost penalty [Burman 10] to counter ill cuts (gradient jump penalty across faces near curved boundary/interface)

## **Unfitted HHO**

- Main ideas [Burman, AE 18 (SINUM)]
  - double cell and face dofs in cut cells, no dofs on curved boundary/interface
  - mixed-order setting:  $k \ge 0$  for face dofs and (k + 1) for cell dofs
  - local cell agglomeration as an alternative to ghost penalty see [Sollie, Bokhove, van der Vegt 11; Johansson, Larson 13] for dG context

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- Improvements in [Burman, Cicuttin, Delay, AE 21 (SISC)]
  - novel gradient reconstruction  $\Rightarrow O(1)$  penalty parameter
  - robust cell agglomeration procedure (ensures locality)
- Extensions
  - Stokes interface problems [Burman, Delay, AE 20 (IMANUM)]
  - wave propagation [Burman, Duran, AE 21 (CMAME)]

Global dofs



$$\hat{u}_h \in \hat{U}_h := \bigotimes_{T \in \mathcal{T}^1} \mathbb{P}^{k+1}(T_1) \times \bigotimes_{T \in \mathcal{T}^2} \mathbb{P}^{k+1}(T_2) \times \bigotimes_{F \in \mathcal{F}^1} \mathbb{P}^k(F_1) \times \bigotimes_{F \in \mathcal{F}^2} \mathbb{P}^k(F_2)$$

- $\bullet\,$  We set to zero all the face components attached to  $\partial\Omega$
- All the cell dofs are eliminated locally by static condensation
- Only the face dofs are globally coupled

## Illustration of agglomeration procedure

#### • Circular interface



• Flower-like interface



## Links to other methods

## HHO $\equiv$ WG $\equiv$ HDG $\equiv$ ncVEM

- [Cockburn, Di Pietro, AE 16 (M2AN)], [Di Pietro, Droniou, Manzini 18 (JCP)], [Cicuttin, AE, Pignet 21 (SpringerBriefs)]
- !! Different devising viewpoints should be mutually enriching !!

## Weak Galerkin (WG)

- WG methods devised in [Wang, Ye 13] (vast litterature...)
- Similar devising of HHO and WG
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- HHO gradient reconstruction is called weak gradient in WG
- WG often uses plain LS stabilization

$$S_{\partial T}^{\rm wG}(\delta_{\hat{u}_T}) := \delta_{\hat{u}_T} \quad \text{vs.} \quad S_{\partial T}^{\rm HHO}(\delta_{\hat{u}_T}) := \begin{cases} \Pi_{\partial T}^k (\delta_{\hat{u}_T} - ((I - \Pi_T^k)R_T(0, \delta_{\hat{u}_T}))|_{\partial T}) & (l = k) \\ \Pi_{\partial T}^k (\delta_{\hat{u}_T}) & (l = k+1) \end{cases}$$

Plain LS stabilization leads to O(h<sup>k</sup>) H<sup>1</sup>-error bounds (not O(h<sup>k+1</sup>) ...)
 achieving O(h<sup>k+1</sup>) bounds requires using face polynomials of order (k + 1) ⇒ more expensive

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  - the local equation for the dual variable is the grad. rec. formula in HHO!
  - one passes from HDG to HHO formulation by static condensation of dual variable

$$\begin{bmatrix} \mathsf{A}_{\sigma\sigma}^{\mathsf{HDG}} & \mathsf{A}_{\sigma\iota\iota}^{\mathsf{HDG}} & \mathsf{A}_{\sigma\iota\iota}^{\mathsf{HDG}} \\ \mathsf{A}_{u\sigma}^{\mathsf{HDG}} & \mathsf{A}_{uu}^{\mathsf{HDG}} & \mathsf{A}_{u\iota\iota}^{\mathsf{HDG}} \\ \mathsf{A}_{u\sigma}^{\mathsf{HDG}} & \mathsf{A}_{\lambda u}^{\mathsf{HDG}} & \mathsf{A}_{\lambda\iota\iota}^{\mathsf{HDG}} \end{bmatrix} \begin{bmatrix} \mathsf{S}_{\mathcal{T}} \\ \mathsf{U}_{\mathcal{T}} \\ \mathsf{U}_{\mathcal{F}} \end{bmatrix} = \begin{bmatrix} \mathsf{0} \\ \mathsf{F}_{\mathcal{T}} \\ \mathsf{0} \end{bmatrix} \qquad \Longleftrightarrow \qquad \begin{cases} \mathsf{A}_{\sigma\sigma}^{\mathsf{HDG}} \mathsf{S}_{\mathcal{T}} = -(\mathsf{A}_{\sigma\iota\iota}^{\mathsf{HDG}} \mathsf{U}_{\mathcal{T}} + \mathsf{A}_{\sigma\iota\iota}^{\mathsf{HDG}} \mathsf{U}_{\mathcal{F}}) \\ \left[ \mathsf{A}_{uu}^{\mathsf{HHO}} & \mathsf{A}_{u\iota\iota}^{\mathsf{HHO}} \\ \mathsf{A}_{u\iota\iota}^{\mathsf{HHO}} & \mathsf{A}_{\iota\iota\iota}^{\mathsf{HHO}} \\ \mathsf{A}_{\lambda\iota\iota}^{\mathsf{HHO}} & \mathsf{A}_{\lambda\iota\iota}^{\mathsf{HHO}} \end{bmatrix} \begin{bmatrix} \mathsf{U}_{\mathcal{T}} \\ \mathsf{U}_{\mathcal{F}} \end{bmatrix} = \begin{bmatrix} \mathsf{F}_{\mathcal{T}} \\ \mathsf{0} \end{bmatrix}$$

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- HHO is an HDG method!
  - this bridge uncovers HHO numerical flux trace

$$\widehat{\mathbf{q}}_{\partial T}(\widehat{u}_T) = -\mathbf{G}_T(\widehat{u}_T) \cdot \mathbf{n}_T + h_T^{-1}(S_{\partial T}^{\star} \circ S_{\partial T})(\delta_{\widehat{u}_T})$$

- one HHO novelty: use of reconstruction in stabilization (equal-order case)
- Main HHO benefit: simpler analysis based on *L*<sup>2</sup>-projections (avoids special HDG projection!)
- ncVEM devised in [Ayuso, Manzini, Lipnikov 16]
- Virtual space

$$\mathbb{P}^{k+1}(T) \subsetneq \mathcal{V}_T := \{ v \in H^1(T) \mid \Delta v \in \mathbb{P}^l(T), \ \mathbf{n} \cdot \nabla v |_{\partial T} \in \mathbb{P}^k(\mathcal{F}_{\partial T}) \}$$

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- HHO dof space  $\hat{U}_T$  with l := k 1 isomorphic to virtual space  $\mathcal{V}_T$ 
  - virtual reconstruction operator  $\mathcal{R}_T : \hat{U}_T \to \mathcal{V}_T$
  - $\hat{\mathcal{J}}_T : \mathcal{V}_T \to \hat{U}_T$ : restriction of reduction operator to virtual space
  - then,  $\hat{\mathcal{J}}_T \circ \mathcal{R}_T = I_{\hat{\mathcal{U}}_T}$  and  $\mathcal{R}_T \circ \hat{\mathcal{J}}_T = I_{\mathcal{V}_T}$

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- Further link to Multiscale Hybrid Mixed (MHM methods) [Chaumont, AE, Lemaire, Valentin 22]

# Wave propagation problems

- Second-order formulation in time: Newmark schemes
- First-order formulation in time: Runge-Kutta (RK) schemes
- [Burman, Duran, AE 22 (CAMC, CMAME)], [Burman, Duran, AE, Steins 21 (JSC)], [Steins, AE, Jamond, Drui 23 (M2AN)]

- Domain  $\Omega \subset \mathbb{R}^d$ , time interval  $J := (0, T_f), T_f > 0$
- Acoustic wave equation with wave speed  $c := \sqrt{\kappa/\rho}$

$$(\partial_{tt}p(t),w)_{\frac{1}{\kappa};\Omega} + (\nabla p(t),\nabla w)_{\frac{1}{\rho};\Omega} = (f(t),w)_{\Omega}, \quad \forall w \in H^1_0(\Omega) \ \forall t \in J$$

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• Energy balance:  $\mathfrak{E}(t) = \mathfrak{E}(0) + \int_0^t (f(s), \partial_t p(s))_{\Omega} ds$  with

$$\mathfrak{E}(t) := \frac{1}{2} \|\partial_t p(t)\|_{\frac{1}{\kappa};\Omega}^2 + \frac{1}{2} \|\nabla p(t)\|_{\frac{1}{\rho};\Omega}^2$$

- Domain  $\Omega \subset \mathbb{R}^d$ , time interval  $J := (0, T_f), T_f > 0$
- Acoustic wave equation with wave speed  $c := \sqrt{\kappa/\rho}$

$$(\partial_{tt}p(t),w)_{\frac{1}{\kappa};\Omega} + (\nabla p(t),\nabla w)_{\frac{1}{\rho};\Omega} = (f(t),w)_{\Omega}, \quad \forall w \in H^1_0(\Omega) \; \forall t \in J$$

• Energy balance:  $\mathfrak{E}(t) = \mathfrak{E}(0) + \int_0^t (f(s), \partial_t p(s))_{\Omega} ds$  with

$$\mathfrak{E}(t) := \frac{1}{2} \|\partial_t p(t)\|_{\frac{1}{\kappa};\Omega}^2 + \frac{1}{2} \|\nabla p(t)\|_{\frac{1}{\rho};\Omega}^2$$

• Everything can be extended to elastodynamics

#### Recap on HHO tools

• Local cell dofs in  $\mathbb{P}^{l}(T)$ ,  $l \in \{k, k+1\}$ , and local face dofs in  $\mathbb{P}^{k}(\mathcal{F}_{\partial T})$ 

$$\hat{u}_T = (u_T, u_{\partial T}) \in \hat{U}_T := \mathbb{P}^l(T) \times \mathbb{P}^k(\mathcal{F}_{\partial T})$$

- Local gradient reconstruction  $\mathbf{G}_T(\hat{u}_T) \in \mathbb{P}^k(T; \mathbb{R}^d)$  (or in  $\nabla \mathbb{P}^{k+1}(T)$ )
- Local stabilization acting on  $\delta_{\hat{u}_T} := u_T |_{\partial T} u_{\partial T}$

$$S_{\partial T}(\delta_{\hat{u}_T}) := \begin{cases} \Pi_{\partial T}^k (\delta_{\hat{u}_T} - \left( (I - \Pi_T^k) R_T(0, \delta_{\hat{u}_T}) \right) |_{\partial T} ) & \text{if } l = k \\ \Pi_{\partial T}^k (\delta_{\hat{u}_T}) & \text{if } l = k+1 \end{cases}$$

• Local bilinear form (with  $\tau_{\partial T} := (\rho_{|T}h_T)^{-1}$ )

$$a_T(\hat{u}_T, \hat{w}_T) := (\mathbf{G}_T(\hat{u}_T), \mathbf{G}_T(\hat{w}_T))_{\frac{1}{\rho}; T} + \tau_{\partial T}(S_{\partial T}(\delta_{\hat{u}_T}), S_{\partial T}(\delta_{\hat{w}_T}))_{\partial T}$$

• Global bilinear form  $a_h$  on HHO space  $\hat{U}_{h0}$  (with Dirichlet BCs)

### HHO space semi-discretization

• Space semi-discrete form: Find  $\hat{p}_h \in C^2(\overline{J}; \hat{U}_{h0})$  s.t.

$$(\partial_{tt} p_{\mathcal{T}}(t), w_{\mathcal{T}})_{\frac{1}{\kappa};\Omega} + a_h(\hat{p}_h(t), \hat{w}_h) = (f(t), w_{\mathcal{T}})_{\Omega}, \quad \forall \hat{w}_h \in \hat{U}_{h0} \, \forall t \in J$$

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• Energy balance:  $\mathfrak{E}_h(t) = \mathfrak{E}_h(0) + \int_0^t (f(s), \partial_t p_{\mathcal{T}}(s))_{\Omega} ds$  with

$$\mathfrak{E}_{h}(t) := \frac{1}{2} \|\partial_{t} p_{\mathcal{T}}(t)\|_{\frac{1}{k};\Omega}^{2} + \frac{1}{2} \|\mathbf{G}_{\mathcal{T}}(\hat{p}_{h}(t))\|_{\frac{1}{\rho};\Omega}^{2} + \frac{1}{2} s_{h}(\hat{p}_{h}(t), \hat{p}_{h}(t))$$

Stabilization is taken into account in the energy definition

• HDG methods for wave equation in second-order form [Cockburn, Fu, Hungria, Ji, Sanchez, Sayas 18]

• Bases for  $\mathbb{P}^{l}(\mathcal{T})$  and  $\mathbb{P}^{k}(\mathcal{F}) \Longrightarrow$  vectors  $(\mathsf{P}_{\mathcal{T}}(t), \mathsf{P}_{\mathcal{F}}(t)) \in \mathbb{R}^{N_{\mathcal{T}}} \times \mathbb{R}^{N_{\mathcal{F}}}$ 

$$\begin{bmatrix} \mathsf{M}_{\mathcal{T}\mathcal{T}}\partial_{tt}\mathsf{P}_{\mathcal{T}}(t)\\ 0 \end{bmatrix} + \begin{bmatrix} \mathsf{A}_{\mathcal{T}\mathcal{T}} & \mathsf{A}_{\mathcal{T}\mathcal{F}}\\ \mathsf{A}_{\mathcal{F}\mathcal{T}} & \mathsf{A}_{\mathcal{F}\mathcal{F}} \end{bmatrix} \begin{bmatrix} \mathsf{P}_{\mathcal{T}}(t)\\ \mathsf{P}_{\mathcal{F}}(t) \end{bmatrix} = \begin{bmatrix} \mathsf{F}_{\mathcal{T}}(t)\\ 0 \end{bmatrix}$$

- $\bullet\,$  Mass matrix  $M_{\mathcal{TT}}$  and stiffness submatrix  $A_{\mathcal{TT}}$  are block-diagonal
- Stiffness submatrix A<sub>FF</sub> is only sparse: face dofs from the same cell are coupled together owing to reconstruction

- [Burman, Duran, AE, Steins 21 (JSC)] proves (for smooth solutions)
  - $\|\partial_t p \partial_t p_{\mathcal{T}}\|_{L^{\infty}(J;L^2(\frac{1}{k};\Omega))} + \|\nabla p \mathbf{G}_{\mathcal{T}}(\hat{p}_h)\|_{L^2(J;L^2(\frac{1}{p};\Omega))} \lesssim h^{k+1}$   $\|\Pi^l_{\mathcal{T}}(p) p_{\mathcal{T}}\|_{L^{\infty}(J;L^2(\frac{1}{\alpha};\Omega))} \lesssim h^{k+2}$  (under full elliptic regularity)
- Some comments on proofs
  - adapt ideas from FEM analysis [Dupont 73; Wheeler 73; Baker 76]
  - simpler than HDG (which needs special initialization)
  - applies to DG using discr. gradients (revisit [Grote, Schneebeli, Schötzau 06])

#### Newmark schemes

- Newmark scheme with parameters  $(\beta, \gamma) = (\frac{1}{4}, \frac{1}{2})$ 
  - implicit, second-order, unconditionally stable
  - $p, \partial_t p, \partial_{tt} p$  are approximated by hybrid pairs  $\hat{p}_h^n, \hat{v}_h^n, \hat{a}_h^n \in \hat{U}_{h0}, \forall n \ge 0$

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#### Newmark schemes

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- Discrete energy is exactly conserved
- Improvements on leapfrog scheme [Steins, AE, Jamond, Drui 23 (M2AN)]
  - $\bullet\,$  plain leapfrog not efficient: needs inverting stiffness submatrix  $A_{\mathcal{FF}}$
  - one can use an iterative method exploiting bock-diagonal structure of face-face penalty submatrix
  - convergence guaranteed if stabilization scaled with large enough weight
    - sharp estimate depending on trace inequality constant (*h*-independent)
    - mild impact on CFL condition despite increased stiffness (up to factor of 2)
  - computational performances
    - close-to-optimal value of weight easy to set
    - generally outperforms plain leapfrog, especially for nonlinear problems
    - mixed-order HHO setting more efficient than equal-order

#### First-order formulation in time

• Introduce velocity  $v := \partial_t p$  and dual variable  $\sigma := \frac{1}{\rho} \nabla p$ 

• Weak form:  $\forall (\tau, w) \in L^2(\Omega; \mathbb{R}^d) \times H^1_0(\Omega), \forall t \in J$ ,

$$\begin{cases} (\partial_t \boldsymbol{\sigma}(t), \boldsymbol{\tau})_{\rho;\Omega} - (\nabla \boldsymbol{v}(t), \boldsymbol{\tau})_{\Omega} = 0 & \leftrightarrow & \rho \partial_t \boldsymbol{\sigma} - \nabla \boldsymbol{v} = 0 \\ (\partial_t \boldsymbol{v}(t), \boldsymbol{w})_{\frac{1}{\kappa};\Omega} + (\boldsymbol{\sigma}(t), \nabla \boldsymbol{w})_{\Omega} = (f(t), \boldsymbol{w})_{\Omega} & \leftrightarrow & \frac{1}{\kappa} \partial_t \boldsymbol{v} - \operatorname{div} \boldsymbol{\sigma} = f \end{cases}$$

• Energy balance:  $\mathfrak{E}(t) = \mathfrak{E}(0) + \int_0^t (f(s), v(s))_{\Omega} ds$  with

$$\mathfrak{E}(t) := \frac{1}{2} \| \boldsymbol{v}(t) \|_{\frac{1}{\kappa};\Omega}^2 + \frac{1}{2} \| \boldsymbol{\sigma}(t) \|_{\rho;\Omega}^2$$

#### HHO space semi-discretization

- $\hat{v}_h \in C^1(\overline{J}; \hat{U}_{h0})$  and  $\sigma_{\mathcal{T}} \in C^1(\overline{J}; \mathbf{S}_{\mathcal{T}})$  with  $\mathbf{S}_{\mathcal{T}} := \mathbb{P}^k(\mathcal{T}; \mathbb{R}^d)$
- Space semi-discrete form:

$$\begin{aligned} & \left( (\partial_t \boldsymbol{\sigma}_{\mathcal{T}}(t), \boldsymbol{\tau}_{\mathcal{T}})_{\rho;\Omega} - (\mathbf{G}_{\mathcal{T}}(\hat{v}_h(t)), \boldsymbol{\tau}_{\mathcal{T}})_{\Omega} = 0 \\ & (\partial_t v_{\mathcal{T}}(t), w_{\mathcal{T}})_{\frac{1}{\kappa};\Omega} + (\boldsymbol{\sigma}_{\mathcal{T}}(t), \mathbf{G}_{\mathcal{T}}(\hat{w}_h))_{\Omega} + \tilde{s}_h(\hat{v}_h(t), \hat{w}_h) = (f(t), w_{\mathcal{T}})_{\Omega} \end{aligned} \right) \end{aligned}$$

• Stabilization  $\tilde{s}_h(\cdot, \cdot)$  with weight  $\tilde{\tau}_{\partial T} = O(h_T^{-\alpha})$ , one takes  $\alpha \in \{0, 1\}$ 

#### HHO space semi-discretization

- $\hat{v}_h \in C^1(\overline{J}; \hat{U}_{h0})$  and  $\sigma_{\mathcal{T}} \in C^1(\overline{J}; \mathbf{S}_{\mathcal{T}})$  with  $\mathbf{S}_{\mathcal{T}} := \mathbb{P}^k(\mathcal{T}; \mathbb{R}^d)$
- Space semi-discrete form:

$$\begin{aligned} \left\{ (\partial_t \boldsymbol{\sigma}_{\mathcal{T}}(t), \boldsymbol{\tau}_{\mathcal{T}})_{\rho;\Omega} - (\mathbf{G}_{\mathcal{T}}(\hat{v}_h(t)), \boldsymbol{\tau}_{\mathcal{T}})_{\Omega} &= 0 \\ (\partial_t \boldsymbol{v}_{\mathcal{T}}(t), \boldsymbol{w}_{\mathcal{T}})_{\frac{1}{\kappa};\Omega} + (\boldsymbol{\sigma}_{\mathcal{T}}(t), \mathbf{G}_{\mathcal{T}}(\hat{w}_h))_{\Omega} + \tilde{s}_h(\hat{v}_h(t), \hat{w}_h) &= (f(t), \boldsymbol{w}_{\mathcal{T}})_{\Omega} \end{aligned} \right.$$

- Stabilization  $\tilde{s}_h(\cdot, \cdot)$  with weight  $\tilde{\tau}_{\partial T} = O(h_T^{-\alpha})$ , one takes  $\alpha \in \{0, 1\}$
- Energy balance:  $\mathfrak{E}_h(t) := \frac{1}{2} \| v_{\mathcal{T}}(t) \|_{\frac{1}{2};\Omega}^2 + \frac{1}{2} \| \boldsymbol{\sigma}_{\mathcal{T}}(t) \|_{\rho;\Omega}^2$

$$\mathfrak{E}_h(t) + \int_0^t \tilde{s}_h(\hat{v}_h(s), \hat{v}_h(s)) ds = \mathfrak{E}_h(0) + \int_0^t (f(s), \mathbf{v}_{\mathcal{T}}(s))_\Omega ds$$

Stabilization acts as a dissipative mechanism

• HDG methods for wave equation in first-order form [Nguyen, Peraire, Cockburn 11; Stranglmeier, Nguyen, Peraire, Cockburn 16]

#### Algebraic realization

• Component vectors  $Z_{\mathcal{T}}(t) \in \mathbb{R}^{M_{\mathcal{T}}}$  and  $(V_{\mathcal{T}}(t), V_{\mathcal{F}}(t)) \in \mathbb{R}^{N_{\mathcal{T}} \times N_{\mathcal{F}}}$ 

$$\begin{bmatrix} \mathsf{M}^{\boldsymbol{\sigma}}_{\mathcal{T}\mathcal{T}}\partial_{t}\mathsf{Z}_{\mathcal{T}}(t)\\ \mathsf{M}_{\mathcal{T}\mathcal{T}}\partial_{t}\mathsf{V}_{\mathcal{T}}(t)\\ 0 \end{bmatrix} + \begin{bmatrix} 0 & -\mathsf{G}_{\mathcal{T}} & -\mathsf{G}_{\mathcal{F}}\\ \mathsf{G}^{\dagger}_{\mathcal{T}} & \mathsf{S}_{\mathcal{T}\mathcal{T}} & \mathsf{S}_{\mathcal{T}\mathcal{F}}\\ \mathsf{G}^{\dagger}_{\mathcal{F}} & \mathsf{S}_{\mathcal{T}\mathcal{T}} & \mathsf{S}_{\mathcal{T}\mathcal{F}}\\ \mathsf{G}^{\dagger}_{\mathcal{F}} & \mathsf{S}_{\mathcal{F}\mathcal{T}} & \mathsf{S}_{\mathcal{F}\mathcal{F}} \end{bmatrix} \begin{bmatrix} \mathsf{Z}_{\mathcal{T}}(t)\\ \mathsf{V}_{\mathcal{T}}(t)\\ \mathsf{V}_{\mathcal{T}}(t)\\ \mathsf{V}_{\mathcal{F}}(t) \end{bmatrix} = \begin{bmatrix} 0\\ \mathsf{F}_{\mathcal{T}}(t)\\ 0 \end{bmatrix}$$

- Mass matrices  $M^{\sigma}_{TT}$  and  $M_{TT}$  are block-diagonal
- Key point: stab. submatrix  $S_{\mathcal{FF}}$  block-diagonal only if l = k + 1
  - for *l* = *k*, high-order HHO correction in stabilization destroys this property (couples all faces of the same cell)
  - mixed-order HHO setting recommended for explicit schemes!

- Natural choice for first-order formulation in time
  - single diagonally implicit RK: SDIRK(*s*, *s* + 1) (*s* stages, order (*s* + 1))
  - explicit RK: ERK(s) (s stages, order s)
- ERK schemes subject to CFL stability condition  $\frac{c\Delta t}{h} \leq \beta(s)\mu(k)$ 
  - $\beta(s)$  slightly increases with  $s \in \{2, 3, 4\}$
  - $\mu(k)$  essentially behaves as  $(k + 1)^{-1}$  w.r.t. polynomial degree

#### 1D heterogeneous media

- 1D test case,  $\Omega_1 = (0, 0.5), \Omega_2 = (0.5, 1), c_1/c_2 = 10$ 
  - initial Gaussian profile in  $\Omega_1$
  - analytical solution available (series)
- Benefits of increasing polynomial degree
  - Newmark scheme, equal-order,  $k \in \{1, 2, 3\}, h = 0.1 \times 2^{-8}, \Delta t = 0.1 \times 2^{-9}$
  - HHO-Newmark solution at  $t = \frac{1}{2}$  (after reflection/transmission at  $x = \frac{1}{2}$ )



#### 2D heterogeneous media

• 2D test case, Ricker (Mexican hat) wavelet

• 
$$\Omega_1 = (0,1) \times (0,\frac{1}{2}), \Omega_2 = (0,1) \times (\frac{1}{2},1), c_1/c_2 = 5$$

• 
$$p_0 = 0, v_0 = -\frac{4}{10}\sqrt{\frac{10}{3}}\left(1600 r^2 - 1\right)\pi^{-\frac{1}{4}}\exp\left(-800r^2\right),$$
  
 $r^2 = (x - x_c)^2 + (y - y_c)^2, (x_c, y_c) = (\frac{1}{2}, \frac{1}{4}) \in \Omega_1$ 

• semi-analytical solution (infinite media): gar6more2d software (INRIA)

#### • HHO-SDIRK(3,4) velocity profiles

- mixed-order, k = 5, polygonal meshes
- $\Delta t = 0.025 \times 2^{-6}$  (four times larger than Newmark for similar accuracy)



### Wave propagation across interface

- Subdomains  $\Omega_1, \Omega_2 \subset \Omega$ , interface  $\Gamma$ , jump  $\llbracket a \rrbracket_{\Gamma} = a_{|\Omega_1} a_{|\Omega_2}$
- Acoustic wave propagation across interface

$$\begin{cases} \frac{1}{\kappa} \partial_{tt} p - \operatorname{div} \left( \frac{1}{\rho} \nabla p \right) = f & \text{in } J \times (\Omega_1 \cup \Omega_2) \\ \llbracket p \rrbracket_{\Gamma} = 0, \ \llbracket \frac{1}{\rho} \nabla p \rrbracket_{\Gamma} \cdot \mathbf{n}_{\Gamma} = 0 & \text{on } J \times \Gamma \end{cases}$$

- Use main ideas from elliptic interface problems
  - mixed-order setting l = k + 1
  - distinct gradient reconstructions  $\mathbf{G}_{T_i}$  in  $\mathbb{P}^k(T_i; \mathbb{R}^d), i \in \{1, 2\}$
  - O(1) penalty parameter
  - LS stabilization on  $(\partial T)^i$ ,  $i \in \{1, 2\} \Longrightarrow s_{T_i}(\cdot, \cdot)$
- Unfitted HHO-Newmark, ERK and SDIRK available

#### Fitted-unfitted comparison

• 2D heterogeneous test case with flat interface

• 
$$\Omega_1 := (-\frac{3}{2}, \frac{3}{2}) \times (-\frac{3}{2}, 0), \Omega_2 := (-\frac{3}{2}, \frac{3}{2}) \times (0, \frac{3}{2})$$

- Ricker wavelet centered at  $(0, \frac{2}{3}) \in \Omega_2$ , sensor  $S_1 = (\frac{3}{4}, -\frac{1}{3}) \in \Omega_1$
- fitted and unfitted HHO behave similarly, both benefit from increasing k

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- fitted and unfitted HHO behave similarly, both benefit from increasing k
- HHO-Newmark,  $\sigma_x$  signals
  - comparison of semi-analytical and HHO (fitted or unfitted) solutions
  - k = 1 (top) and k = 3 (bottom)
  - $c_2/c_1 = \sqrt{3}$  (low contrast, left) or  $c_2/c_1 = 8\sqrt{3}$  (high contrast, right)



## CFL condition for ERK (1/2)

- Homogeneous test case, flat interface
- CFL condition for ERK(s):  $\frac{c\Delta t}{h} \leq \beta(s)\mu(k)$ 
  - $\beta(s)$  mildly depends on the number of stages
  - $\mu(k)$  behaves as  $(k + 1)^{-1}$  and is quantified by solving a generalized eigenvalue problem with the mass and stiffness matrices
- Additional jump penalties in unfitted HHO only mildly impact  $\mu(k)$

k	0	1	2	3
Fitted-HHO	0.118	0.0522	0.0338	0.0229
Unfitted-HHO	0.0765	0.0373	0.0232	0.0159
Ratio	1.5	1.4	1.5	1.4

## CFL condition for ERK (2/2)

- Homogeneous test case, circular interface
  - study of impact of agglomeration parameter  $\theta_{agg}$  on  $\mu(k)$
  - "ill cut" cells flagged if relative area of any subcell falls below  $\theta_{agg}$

• Agglomerated cells for  $\theta_{agg} = 0.3$  on a sequence of refined quad meshes



## CFL condition for ERK (2/2)

0.010 0.005 0.001 5.×10<sup>-4</sup> 1.×10<sup>-4</sup> 5.×10<sup>-5</sup>

- Homogeneous test case, circular interface
  - study of impact of agglomeration parameter  $\theta_{agg}$  on  $\mu(k)$
  - "ill cut" cells flagged if relative area of any subcell falls below  $\theta_{agg}$
- Agglomerated cells for  $\theta_{agg} = 0.3$  on a sequence of refined quad meshes



- Behavior of  $h\mu(k)$  and impact of  $\theta_{agg}$  on  $\mu(k)$ 
  - tolerating ill cut cells deteriorates the CFL condition

	k	0	1	2	3
	$\theta_{\text{agg}} = 0.5$	0.042	0.022	0.014	0.0099
	$\theta_{\text{agg}} = 0.3$	0.030	0.015	0.0094	0.0065
	Ratio	1.4	1.5	1.5	1.5
	$\theta_{agg} = 0.1$	0.017	0.0087	0.0055	0.0039
103	Ratio	2.5	2.6	2.6	2.5
0.01 0.02 0.05 0.10					

## Flower-like interface

• Agglomerated cells for a flower-like interface (quad mesh,  $h = 2^{-5}$ ), HHO-SDIRK(3,4) signal for  $\sigma_x$  at two sensors,  $k \in \{1, 2, 3\}, c_2/c_1 = \sqrt{3}$ 



• Pressure isovalues, SDIRK(3,4), k = 3,  $h = 0.1 \times 2^{-8}$ ,  $\Delta t = 2^{-6}$ 

t = 0.25

vh 0 0.1 2.7e-01 t = 0.5



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• Pressure isovalues, SDIRK(3,4), k = 3,  $h = 0.1 \times 2^{-8}$ ,  $\Delta t = 2^{-6}$ 

t = 0.25

t = 0.5

t = 1



!! Thank you for your attention !!

### Competition: Newmark vs. RK

- All schemes deliver same max. rel. error on a sensor at  $(\frac{1}{2}, \frac{2}{3})$
- Disclaimer: preliminary results! (off-the-shelf solvers)
- If no direct solvers allowed, ERK(4) wins despite CFL restriction
- With direct solvers, SDIRK(3,4) wins
- RK schemes more efficient than Newmark scheme
- for SDIRK(3,4),  $\tilde{\tau}_{\partial T} = O(h_T^{-\alpha})$ ,  $\alpha = 1$  more accurate/expensive than  $\alpha = 0$

scheme	(l,k)	α	solver	t/step	steps	time	err
ERK(4)	(6,5)	0	n/a	0.410	5,120	2,099	2.23
Newmark	(7,6)	1	iter	56.74	2,560	58,265	2.15
SDIRK(3,4)	(6,5)	1	iter	31.24	640	5,639	2.21
SDIRK(3,4)	(6,5)	0	iter	22.52	640	2,200	4.45
Newmark	(7,6)	1	direct	0.515	2,560	1,318	2.15
SDIRK(3,4)	(6,5)	1	direct	1.579	640	1,010	2.21

#### Local dofs



- Mesh still composed of polygonal cells (with planar faces)
- Decomposition of cut cells:  $\overline{T} = \overline{T_1} \cup \overline{T_2}, T^{\Gamma} = T \cap \Gamma$
- Decomposition of cut faces:  $\partial(T_i) = (\partial T)^i \cup T^{\Gamma}, i \in \{1, 2\}$
- Local dofs (no dofs on  $T^{\Gamma}$ !)

 $\hat{u}_T = (u_{T_1}, u_{T_2}, u_{(\partial T)^1}, u_{(\partial T)^2}) \in \mathbb{P}^{k+1}(T_1) \times \mathbb{P}^{k+1}(T_2) \times \mathbb{P}^k(\mathcal{F}_{(\partial T)^1}) \times \mathbb{P}^k(\mathcal{F}_{(\partial T)^2})$ 

#### Gradient reconstruction in cut cells



• Gradient reconstruction  $\mathbf{G}_{T_i}(\hat{u}_T) \in \mathbb{P}^k(T_i; \mathbb{R}^d)$  in each subcell

• (Option 1) Independent reconstruction in each subcell

$$(\mathbf{G}_{T_i}(\hat{u}_T), \mathbf{q})_{T_i} = -(\boldsymbol{u}_{T_i}, \operatorname{div} \mathbf{q})_{T_i} + (\boldsymbol{u}_{(\partial T)^i}, \mathbf{q} \cdot \mathbf{n}_T)_{(\partial T)^i} + (\boldsymbol{u}_{T_i}, \mathbf{q} \cdot \mathbf{n}_{T_i})_{T^{\Gamma}}$$

#### • (Option 2) Reconstruction mixing data from both subcells

 $(\mathbf{G}_{T_i}(\hat{u}_T), \mathbf{q})_{T_i} = -(\mathbf{u}_{T_i}, \operatorname{div} \mathbf{q})_{T_i} + (\mathbf{u}_{(\partial T)^i}, \mathbf{q} \cdot \mathbf{n}_T)_{(\partial T)^i} + (\mathbf{u}_{T_{3-i}}, \mathbf{q} \cdot \mathbf{n}_{T_i})_{T^{\Gamma}}$ 

- Both options avoid Nitsche's consistency terms
  - O(1) penalty parameter

#### Local bilinear form in cut cells

Local bilinear form

$$a_T(\hat{u}_T, \hat{w}_T) := \sum_{i \in \{1, 2\}} \left\{ \kappa_i(\mathbf{G}_{T_i}(\hat{u}_T), \mathbf{G}_{T_i}(\hat{w}_T))_{T_i} + s_{T_i}(\hat{u}_T, \hat{w}_T) \right\} + s_T^{\Gamma}(u_T, w_T)$$

• LS stabilization inside each subdomain

$$s_{T_i}(\hat{u}_T, \hat{w}_T) := \kappa_i h_{T_i}^{-1}(\Pi^k_{(\partial T)^i}(\delta_{\hat{u}_{T_i}}), \delta_{\hat{w}_{T_i}})_{(\partial T)^i}$$

• Interface bilinear form

$$s_T^{\Gamma}(u_T, w_T) := \eta \kappa_1 h_T^{-1}(\llbracket u_T \rrbracket_{\Gamma}, \llbracket w_T \rrbracket_{\Gamma})_{T^{\Gamma}} \text{ with } \eta = O(1)$$

- The use of two gradient reconstructions allows for robustness w.r.t. contrast (κ<sub>1</sub> ≪ κ<sub>2</sub>)
  - use option 1 in  $\Omega_1$  and option 2 in  $\Omega_2$
  - $a_T$  is symmetric, but  $\Omega_1/\Omega_2$  do not play symmetric roles

#### Error analysis

- Multiplicative and discrete trace inequalities [Burman, AE 18]
  - for any cut cell *T*, there is a ball *T*<sup>†</sup> of size *O*(*h<sub>T</sub>*) containing *T* and a finite number of its neighbors, and s.t. all *T* ∩ Γ is visible from a point in *T*<sup>†</sup>
  - small ball with diameter  $O(h_T)$  present on both sides of interface
  - achievable using local cell agglomeration if mesh fine enough

#### Error estimate

Assuming that 
$$u|_{\Omega_i} \in H^{1+t}(\Omega_i)$$
 with  $t \in (\frac{1}{2}, k+1]$ ,

$$\sum_{T} \sum_{i \in \{1,2\}} \kappa_i \|\nabla(u - u_{T_i})\|_{T_i}^2 \le Ch^{2t} \sum_{i \in \{1,2\}} \kappa_i |u|_{H^{t+1}(\Omega_i)}^2$$

Convergence order  $O(h^{k+1})$  if  $u|_{\Omega_i} \in H^{k+2}(\Omega_i)$