

A Posteriori Analysis of a Space-time Hybridizable Discontinuous Galerkin Method for the Advection-diffusion Problem

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Outline

① Motivation and preliminaries

- Advection-diffusion in its space-time formulation
- Advection-dominated regime
- Hybridizable discontinuous Galerkin methods

② A priori error analysis, briefly

- The HDG method
- Péclet-robust inf-sup conditions
- A layer test case

③ A posteriori error analysis

- Why adaptivity
- Reliability, local efficiency and robustness
- Main results
- Two layer test cases

Time-dependent advection-diffusion problem

Standard formulation

$$\partial_t u + \bar{\nabla} \cdot (\bar{\beta} u) - \varepsilon \bar{\nabla}^2 u = f, \quad \text{in } \Omega(t), \quad 0 < t \leq T$$

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- Time derivative: $\partial_t u$

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- Time derivative: $\partial_t u$
- Time-dependent d -dimensional domain: $\Omega(t)$

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- Diffusion: $\varepsilon \bar{\nabla}^2 u$

Time-dependent advection-diffusion problem

Space-time formulation

$$\partial_t u + \bar{\nabla} \cdot (\bar{\beta} u) - \varepsilon \bar{\nabla}^2 u = f, \quad \text{in } \Omega(t), \quad 0 < t \leq T$$



$$\nabla \cdot (\beta u) - \varepsilon \nabla^2 u = f, \quad \text{in } \mathcal{E}$$

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- Space-time advective field: $\beta := (1, \bar{\beta})$
- Space-time gradient: $\nabla := (\partial_t, \bar{\nabla})$
- Space-time advection: $\nabla \cdot (\beta u) = \partial_t u + \bar{\nabla} \cdot (\bar{\beta} u)$

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- Space-time advection: $\nabla \cdot (\beta u) = \partial_t u + \bar{\nabla} \cdot (\bar{\beta} u)$
- $(d + 1)$ -dimensional space-time domain: \mathcal{E}

Solution in the space-time formulation

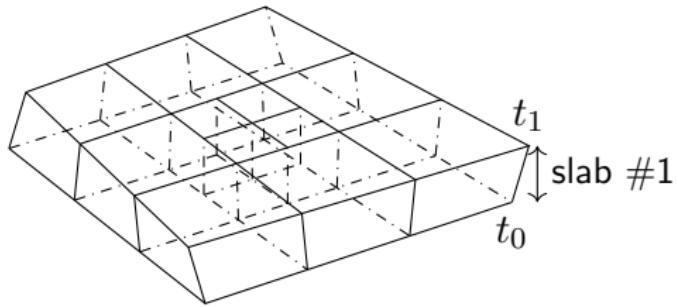
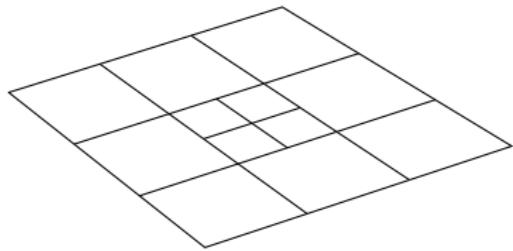
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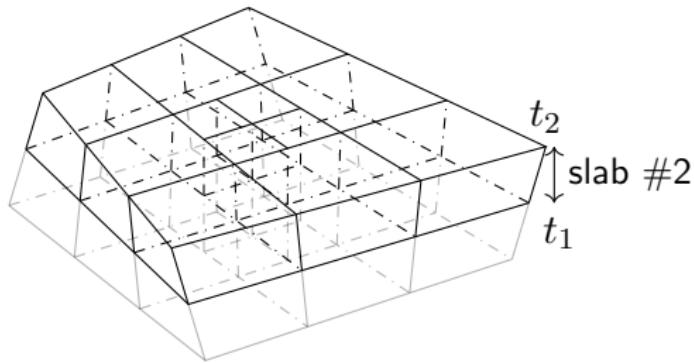
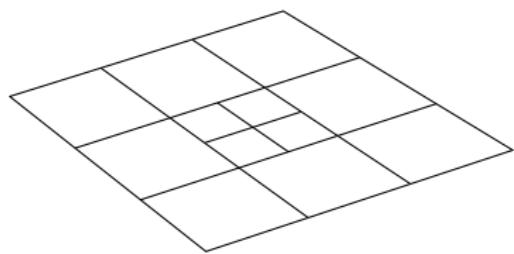
Solution in the space-time formulation

- **All-at-once approach:** discretize in the space-time domain \mathcal{E} all at once
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 - Initial spatial mesh is extruded in time following the mesh deformation
→ space-time slab



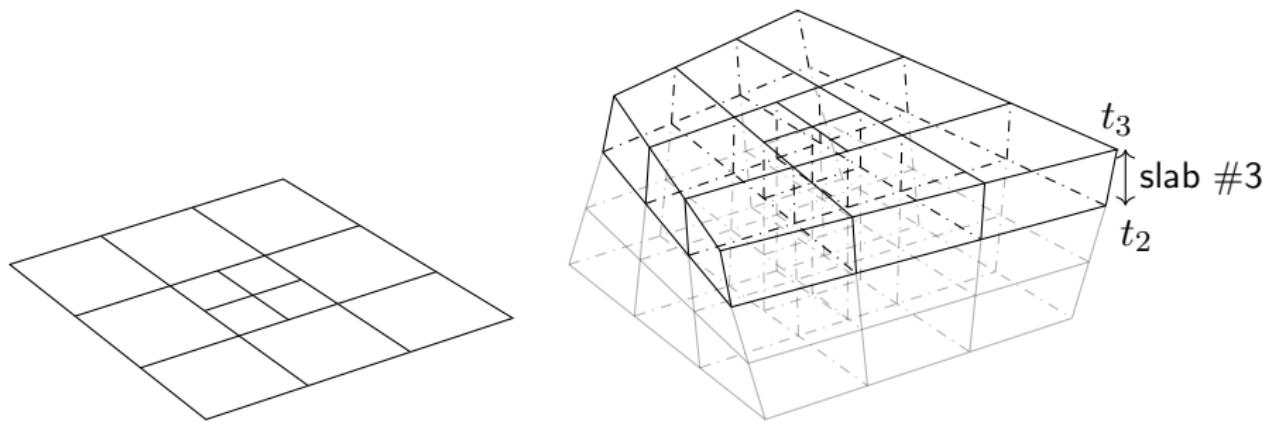
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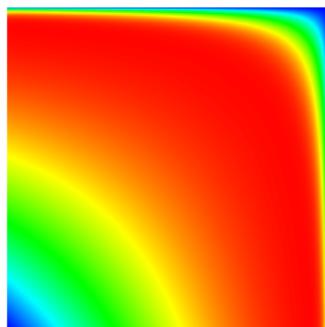
Solution in the space-time formulation

Why space-time

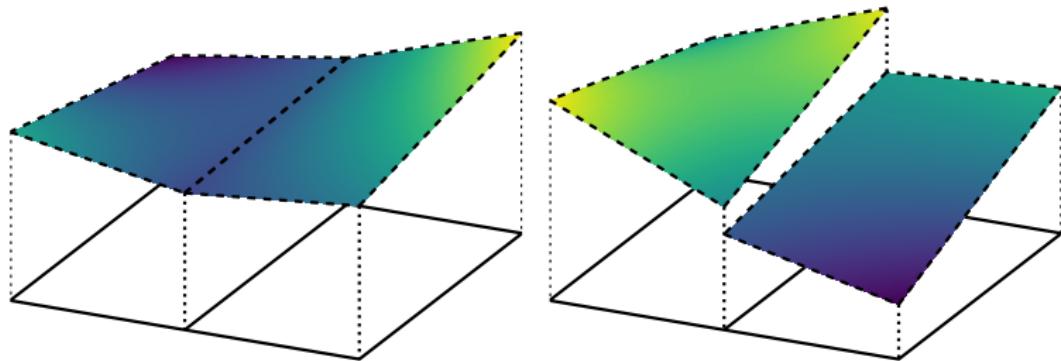
- Built-in domain movement
- Arbitrarily high-order approximation in both space and time
- Local adaptivity in both space and time (more on this later!)

Advection-dominated regime

- The focus of our work
 $0 < \varepsilon \ll 1$ and $\beta \sim \mathcal{O}(1) \Rightarrow \text{P\'eclet number} \gg 1$
- Sharp boundary/interior layers



Discontinuous Galerkin methods



(a) CG: inter-element continuity

(b) DG: discont. piecewise polynomial

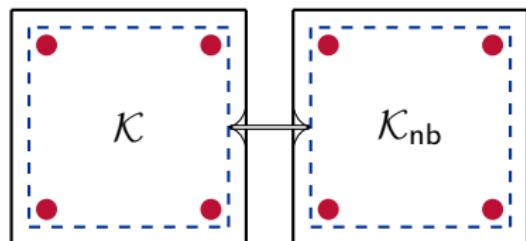
Why DG discretization

- Highly parallelizable implementations
- General meshes (hanging nodes, non-standard shapes)
- hp -adaptivity

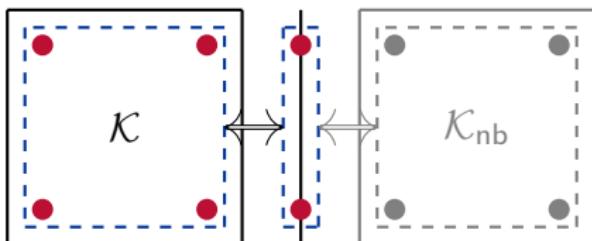
Hybridizable discontinuous Galerkin methods

HDG = DG + Static condensation

- Stabilization properties of DG
- Advantages of DG due to its localized nature
- Reducing the number of globally coupled degrees-of-freedom



(a) DG dofs



(b) HDG dofs

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A space-time HDG discretization

Find $\mathbf{u}_h \in \mathbf{V}_h$ such that

$$a_h(\mathbf{u}_h, \mathbf{v}_h) = (f, v_h)_{\mathcal{T}_h} + \langle g, \mu_h \rangle_{\partial \mathcal{E}_N} \quad \forall \mathbf{v}_h \in \mathbf{V}_h$$

- $a_h(\mathbf{u}_h, \mathbf{v}_h) := a_{h,d}(\mathbf{u}_h, \mathbf{v}_h) + a_{h,c}(\mathbf{u}_h, \mathbf{v}_h)$

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- Diffusion:

$$\begin{aligned} a_{h,d}(\mathbf{u}, \mathbf{v}) := & (\varepsilon \bar{\nabla} u, \bar{\nabla} v)_{\mathcal{T}_h} + \langle \varepsilon \alpha h_K^{-1} [\mathbf{u}], [\mathbf{v}] \rangle_{\mathcal{Q}_h} \\ & - \langle \varepsilon [\mathbf{u}], \bar{\nabla}_{\bar{n}} v \rangle_{\mathcal{Q}_h} - \langle \varepsilon \bar{\nabla}_{\bar{n}} u, [\mathbf{v}] \rangle_{\mathcal{Q}_h} \end{aligned}$$

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- Advection*:

$$\begin{aligned} a_{h,c}(\mathbf{u}, \mathbf{v}) := & -(\beta u, \nabla v)_{\mathcal{T}_h} + \langle \zeta^+ \beta \cdot n \lambda, \mu \rangle_{\partial \mathcal{E}_N} \\ & + \langle (\beta \cdot n) \lambda + \beta_s [\mathbf{u}], [\mathbf{v}] \rangle_{\partial \mathcal{T}_h} \end{aligned}$$

*G. Fu et al. (2015)

The main contribution

Notational setup:

- δt : local time-step
- h : local spatial mesh size
- p_s, p_t are polynomial degrees in spatial and temporal directions

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- δt : local time-step
- h : local spatial mesh size
- p_s, p_t are polynomial degrees in spatial and temporal directions
- The s -norm $\|\cdot\|_s$:

$$\begin{aligned}\|\|v\|\|_v^2 &:= \sum_{\mathcal{K} \in \mathcal{T}_h} \|v\|_{\mathcal{K}}^2 + \sum_{\mathcal{K} \in \mathcal{T}_h} \left\| |\beta_s - \frac{1}{2}\beta \cdot n|^{1/2} [v] \right\|_{\partial\mathcal{K}}^2 + \sum_{F \in \partial\mathcal{E}_N} \left\| |\frac{1}{2}\beta \cdot n|^{1/2} \mu \right\|_F^2 \\ &\quad + \sum_{\mathcal{K} \in \mathcal{T}_h} \varepsilon \|\bar{\nabla} v\|_{\mathcal{K}}^2 + \sum_{\mathcal{K} \in \mathcal{T}_h} \varepsilon h_K^{-1} \|[v]\|_{Q_{\mathcal{K}}}^2 \\ \|\|v\|\|_s^2 &:= \|\|v\|\|_v^2 + \sum_{\mathcal{K} \in \mathcal{T}_h} \tau_{\varepsilon} \|\partial_t v\|_{\mathcal{K}}^2 \quad (\tau_{\varepsilon} = \Delta t_{\mathcal{K}} \tilde{\varepsilon})\end{aligned}$$

The main contribution

A Péclet-robust a priori error analysis

- When $\varepsilon < \delta t = h$

$$\|\| \mathbf{u} - \mathbf{u}_h \| \|_s \leq c_T (h^{p_s+1/2} + \delta t^{p_t+1/2})$$

- When $\delta t = h < \varepsilon$

$$\|\| \mathbf{u} - \mathbf{u}_h \| \|_s \leq c_T (h^{p_s} + \delta t^{p_t})$$

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An $1/2$ drop in the convergence rate is expected after mesh is sufficiently refined

Péclet-robust inf-sup conditions

Main inf-sup conditions:

Norm	Nonrobust	Robust (our contribution)
$\ \cdot\ _v$	$\ w_h\ _v^2 \lesssim \varepsilon^{-1} a_h(w_h, w_h)$	$\Rightarrow \ w_h\ _v \leq c_T \sup_{v_h \in V_h} \frac{a_h(w_h, v_h)}{\ v_h\ _v}$
$\ \cdot\ _s$	$\ w_h\ _s \lesssim \varepsilon^{-1} \sup_{v_h \in V_h} \frac{a_h(w_h, v_h)}{\ v_h\ _s}$	$\Rightarrow \ w_h\ _s \leq c_T \sup_{v_h \in V_h} \frac{a_h(w_h, v_h)}{\ v_h\ _s}$

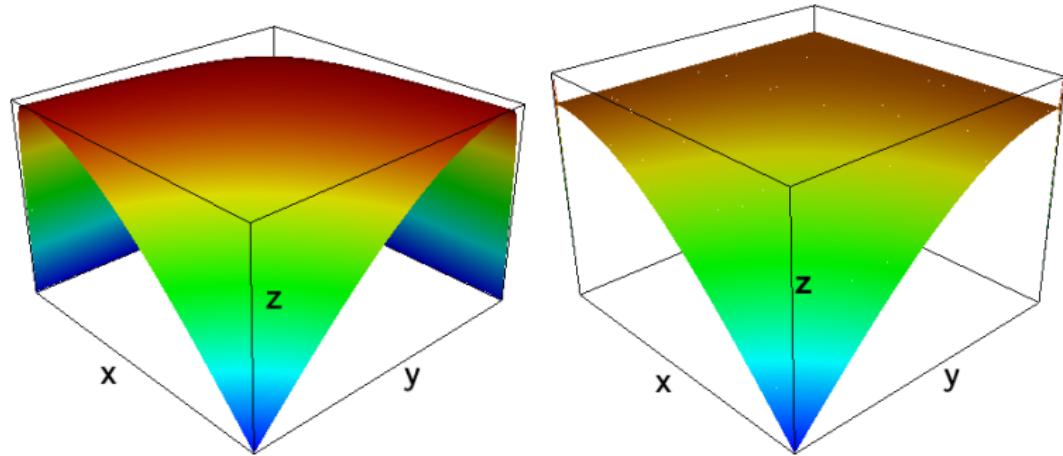
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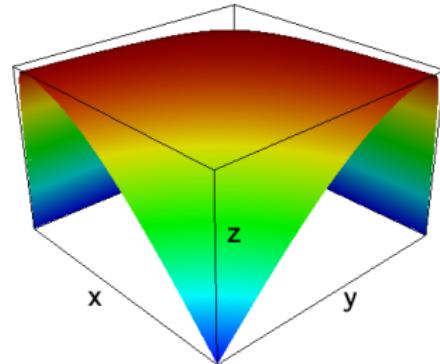
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$\ \cdot\ _s$	$\ w_h\ _s \lesssim \varepsilon^{-1} \sup_{v_h \in V_h} \frac{a_h(w_h, v_h)}{\ v_h\ _s}$	$\ w_h\ _s \leq c_T \sup_{v_h \in V_h} \frac{a_h(w_h, v_h)}{\ v_h\ _s}$

s -norm inf-sup stability proves instrumental in our a posteriori error analysis!

A boundary layer test case



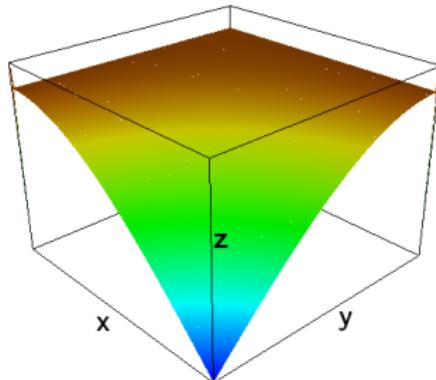
Convergence histories $(\varepsilon = 10^{-2})$



Cells per slab	Slabs	$p = 1$	Rates	$p = 2$	Rates	$p = 3$	Rates
100	10	4.0e-1	-	2.5e-1	-	1.3e-1	-
400	20	2.6e-1	0.6	1.1e-1	1.2	3.3e-2	2.0
1600	40	1.5e-1	0.8	3.5e-2	1.6	6.0e-3	2.5
6400	80	8.0e-2	0.9	9.7e-3	1.9	8.7e-4	2.8
25600	160	4.0e-2	1.0	2.5e-3	2.0	1.1e-4	3.0
102400	320	2.0e-2	1.0	6.2e-4	2.0	1.4e-5	3.0

Convergence histories $(\varepsilon = 10^{-8})$

(error computed outside the layer)



Cells per slab	Slabs	$p = 1$	Rates	$p = 2$	Rates	$p = 3$	Rates
100	10	1.4e-2	-	7.0e-3	-	4.8e-3	-
400	20	1.3e-3	3.3	1.0e-5	9.4	5.3e-8	16.5
1600	40	4.4e-4	1.6	1.6e-6	2.7	3.6e-9	3.9
6400	80	1.5e-4	1.6	2.7e-7	2.6	2.8e-10	3.7
25600	160	5.1e-5	1.5	4.6e-8	2.6	2.3e-11	3.6
102400	320	1.8e-5	1.5	7.9e-9	2.5	2.1e-12	3.5

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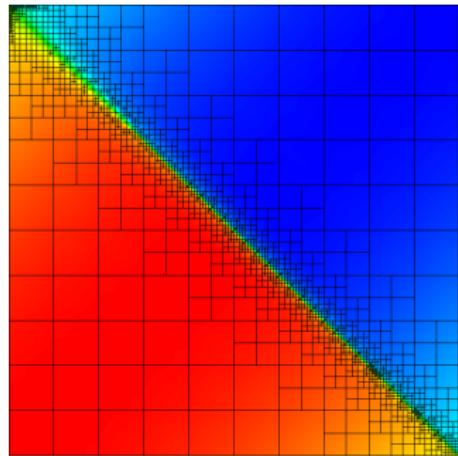
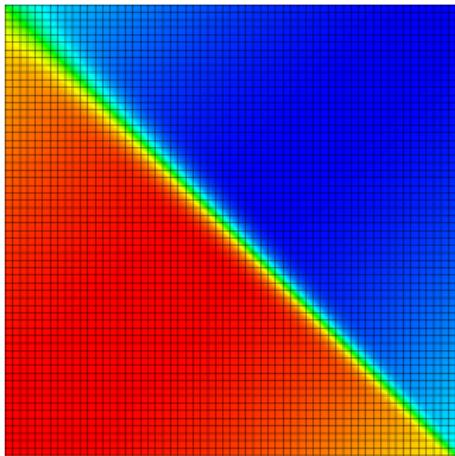
Why adaptivity?

Adaptive mesh refinement (AMR)

- Local errors in the narrow layer region dominate the global error
- Allocate more elements/dofs to areas with larger errors

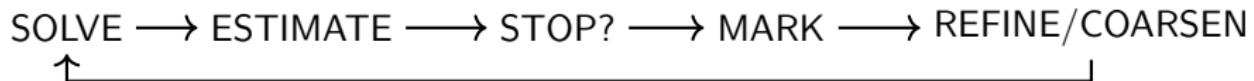
Let's see an example!

A motivating example

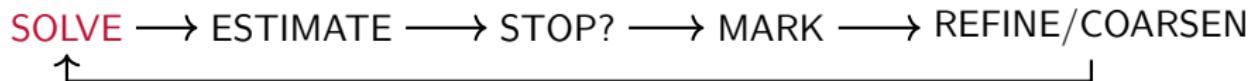


Element	L^2 -error	Order	Element	L^2 -err.	Order
1,000	1.0e-1	-	729	9.3e-5	-
8,000	8.2e-2	0.34	5,832	2.4e-5	2.02
64,000	6.6e-2	0.33	46,656	6.0e-6	2.01
512,000	5.2e-2	0.33	373,248	1.5e-6	2.00

The AMR loop

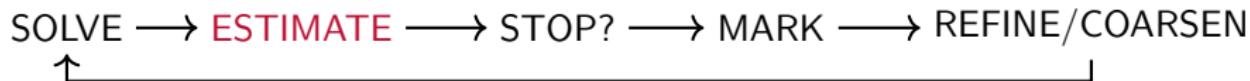


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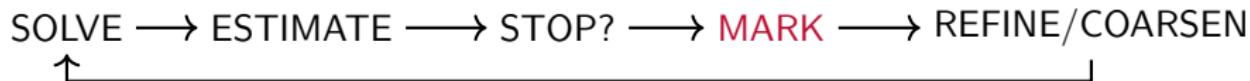


- Given an initial mesh, compute the HDG solution: \mathbf{u}_h
- Compute η_K : the local error estimate on each element
- $(\sum_K \eta_K^2)^{1/2} \leq \tau?$

Reliability:

$$\|u - u_h\|_{\Omega} \leq c^* (\sum_{K \in \mathcal{T}} \eta_K^2)^{1/2} \text{ with } c^* \sim \mathcal{O}(1) \text{ on all refinement levels}$$

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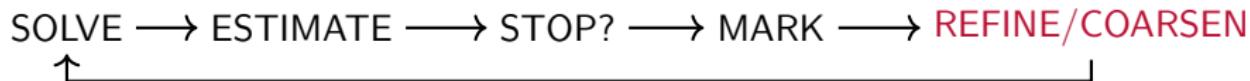
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- Mark elements with η_K bigger than a certain threshold

Local efficiency:

$$\eta_K \leq c_* \|u - u_h\|_K \text{ with } c_* \sim \mathcal{O}(1) \text{ on all refinement levels}$$

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$$\eta_K \leq c_* \|u - u_h\|_K \text{ with } c_* \sim \mathcal{O}(1) \text{ on all refinement levels}$$

- Refine/coarsen marked elements

Main results: the norm and the error estimator

- The error norm being estimated:

$$\begin{aligned}\|v\|_{sT}^2 := & \sum_{\mathcal{K} \in \mathcal{T}_h} \|v\|_{\mathcal{K}}^2 + \sum_{\mathcal{K} \in \mathcal{T}_h} \left\| |\beta_s - \frac{1}{2}\beta \cdot n|^{1/2} [v] \right\|_{\partial\mathcal{K}}^2 + \textcolor{red}{T} \sum_{F \in \partial\mathcal{E}_N} \left\| |\frac{1}{2}\beta \cdot n|^{1/2} \mu \right\|_F^2 \\ & + \textcolor{red}{T} \sum_{\mathcal{K} \in \mathcal{T}_h} \varepsilon \|\bar{\nabla} v\|_{\mathcal{K}}^2 + \sum_{\mathcal{K} \in \mathcal{T}_h} \varepsilon h_K^{-1} \|[v]\|_{Q_{\mathcal{K}}}^2 + \sum_{\mathcal{K} \in \mathcal{T}_h} \tau_{\varepsilon} \|\partial_t v\|_{\mathcal{K}}^2\end{aligned}$$

Main results: the norm and the error estimator

- The error estimator:

$$\eta^2 := \sum_{\mathcal{K} \in \mathcal{T}_h} (\eta^{\mathcal{K}})^2$$

where

$$(\eta^{\mathcal{K}})^2 := (\eta_R^{\mathcal{K}})^2 + \sum_{i=1}^3 (\eta_{J,i}^{\mathcal{K}})^2 + \sum_{j=1}^2 (\eta_{BC,j}^{\mathcal{K}})^2$$

where

$$\eta_R^{\mathcal{K}} := \lambda_{\mathcal{K}} \|R_h^{\mathcal{K}}\|_{\mathcal{K}}$$

$$\eta_{J,1}^{\mathcal{K}} := h_K^{1/2} \varepsilon^{1/2} \|[\bar{\nabla}_{\bar{n}} u_h]\|_{\mathcal{Q}_{\mathcal{K}} \setminus \partial \mathcal{E}}$$

$$\eta_{J,2}^{\mathcal{K}} := ((\eta_{J,2,1}^{\mathcal{K}})^2 + (\eta_{J,2,2}^{\mathcal{K}})^2)^{1/2}$$

$$\eta_{J,3}^{\mathcal{K}} := ((\eta_{J,3,\mathcal{Q}}^{\mathcal{K}})^2 + (\eta_{J,3,\mathcal{R}}^{\mathcal{K}})^2)^{1/2}$$

$$\eta_{BC,1}^{\mathcal{K}} := h_K^{1/2} \varepsilon^{-1/2} \|R_h^N\|_{\mathcal{Q}_{\mathcal{K}} \cap \partial \mathcal{E}_N}$$

$$\eta_{BC,2}^{\mathcal{K}} := \|R_h^N\|_{\partial \mathcal{K} \cap \Omega_0} = \|g - u_h\|_{\partial \mathcal{K} \cap \Omega_0}$$

where

$$\eta_{J,2,1}^{\mathcal{K}} := h_K^{-1/2} \varepsilon^{1/2} \|[\mathbf{u}_h]\|_{\mathcal{Q}_{\mathcal{K}}}$$

$$\eta_{J,2,2}^{\mathcal{K}} := h_K^{1/4} \varepsilon^{-1/2} \|[\mathbf{u}_h]\|_{\mathcal{Q}_{\mathcal{K}}}$$

$$\eta_{J,3,\mathcal{Q}}^{\mathcal{K}} := \||\beta_s - \frac{1}{2}\beta \cdot n|^{1/2} [\mathbf{u}_h]\|_{\mathcal{Q}_{\mathcal{K}}}$$

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Main results: reliability and local efficiency

- Reliability:

$$\|\| \mathbf{u} - \mathbf{u}_h \| \|_{sT} \leq c T \varepsilon^{-1/2} \eta$$

- Local efficiency:

$$\eta^{\mathcal{K}} \leq c \left(\sum_{\mathcal{K} \subset \omega_{\mathcal{K}}} \varepsilon^{-1/2} \tilde{\varepsilon}^{-1/2} \|\| \mathbf{u} - \mathbf{u}_h \| \|_{sT, \mathcal{K}} + \text{osc}_h^{\mathcal{K}} + \text{osc}_h^N \right)$$

- Local efficiency (*when elements are sufficiently refined*):

$$\eta^{\mathcal{K}} \leq c \left(\sum_{\mathcal{K} \subset \omega_{\mathcal{K}}} \varepsilon^{-1/2} \|\| \mathbf{u} - \mathbf{u}_h \| \|_{sT, \mathcal{K}} + \text{osc}_h^{\mathcal{K}} + \text{osc}_h^N \right)$$

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Efficiency index (defined as $\eta / \|\| \mathbf{u} - \mathbf{u}_h \| \|$)

- Pre-asymptotic regime:

bounded below by $\mathcal{O}(\varepsilon^{1/2})$ and above by $\mathcal{O}(\varepsilon^{-1})$;

- Asymptotic regime:

bounded below by $\mathcal{O}(\varepsilon^{1/2})$ and above by $\mathcal{O}(\varepsilon^{-1/2})$;

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Saturation assumption

Halving the time-step of every element of \mathcal{T}_h , obtaining a subgrid mesh $\mathcal{T}_{\mathfrak{h}}$, we assume there exists $\rho < 1$:

$$\sum_{\mathcal{K} \in \mathcal{T}_h} \tau_\varepsilon \|\partial_t(u - u_{\mathfrak{h}})\|_{\mathcal{K}}^2 \leq \rho^2 \sum_{\mathcal{K} \in \mathcal{T}_h} \tau_\varepsilon \|\partial_t(u - u_h)\|_{\mathcal{K}}^2$$

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- The localization of the error estimate is both in space and in time (first to the best of our knowledge)
- However:
 - The saturation assumption technique requires $p_t = 1$
 - The reliability analysis requires $\delta t_K = \mathcal{O}(h_K^2)$
 - The analysis assumes a fixed domain

A boundary layer test case

Exact solution:

$$u(t, x_1, x_2) = (1 - \exp(-t)) \left(\frac{\exp((x_1-1)/\varepsilon)-1}{\exp(-1/\varepsilon)-1} + x_1 - 1 \right) \left(\frac{\exp((x_2-1)/\varepsilon)-1}{\exp(-1/\varepsilon)-1} + x_2 - 1 \right)$$

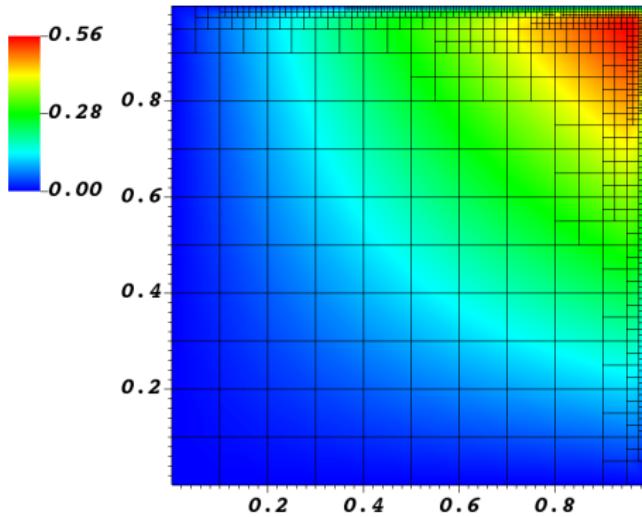


Figure: Solution is for $\varepsilon = 10^{-3}$ and $t = 1.0$

A boundary layer test case

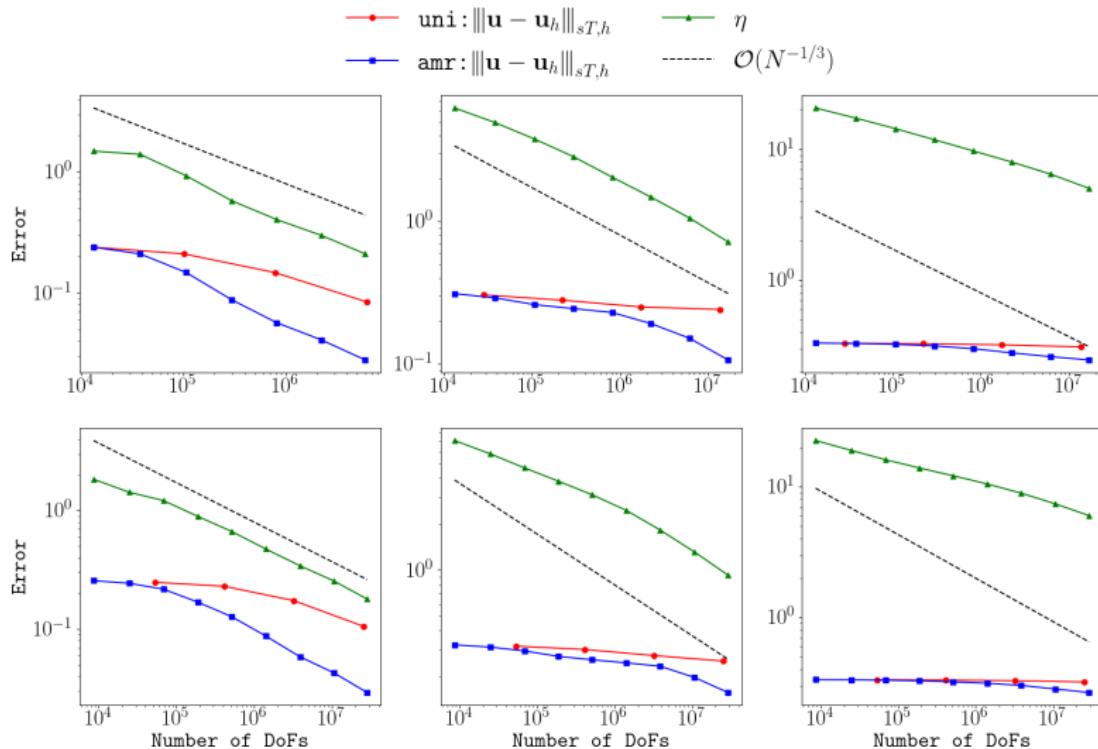
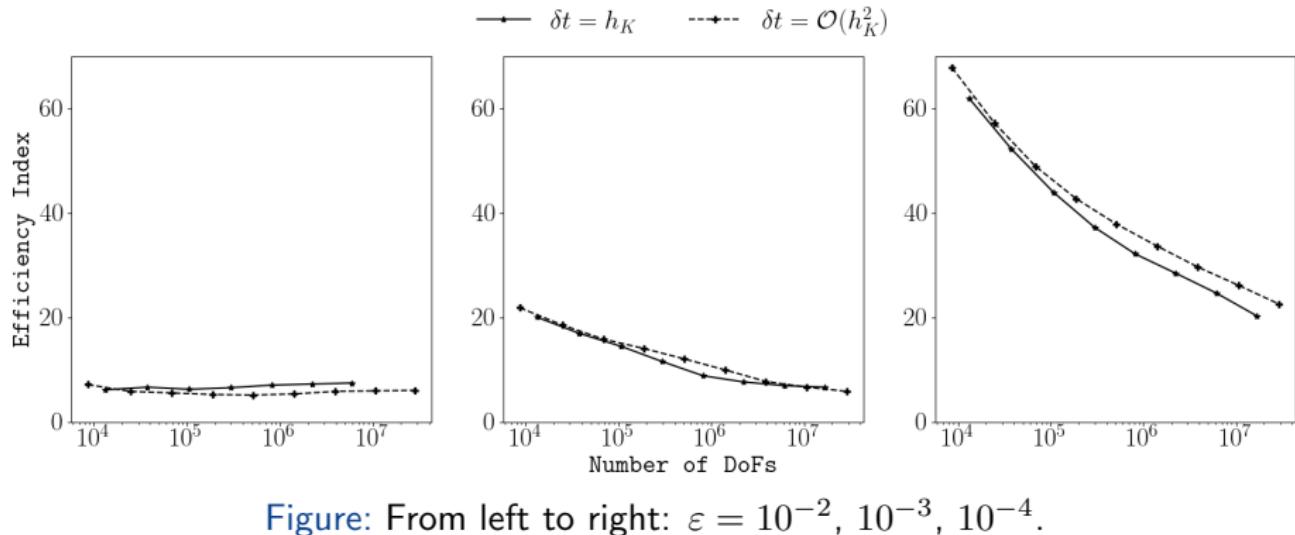


Figure: 1st row: $\delta t_K = h_K$; 2nd row: $\delta t_K = h_K^2$. From left to right: $\epsilon = 10^{-2}, 10^{-3}, 10^{-4}$.

A boundary layer test case



An interior layer test case

Exact solution: $u(t, x_1, x_2) = (1 - \exp(-t)) \left(\arctan\left(\frac{y-x}{\sqrt{2}\varepsilon}\right) \right) \left(1 - \frac{(x+y)^2}{2}\right)$

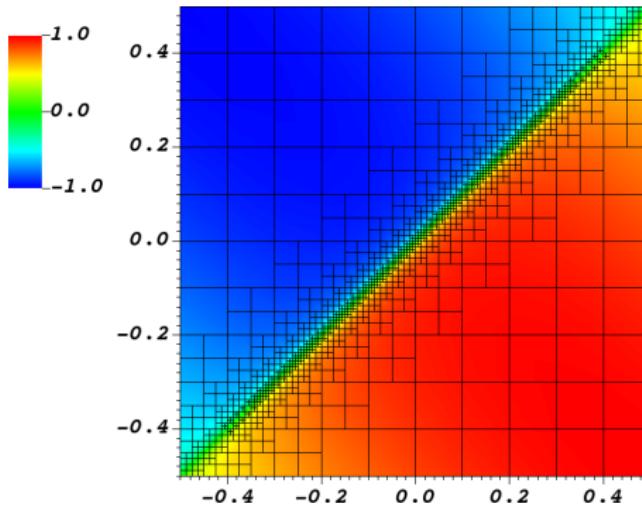


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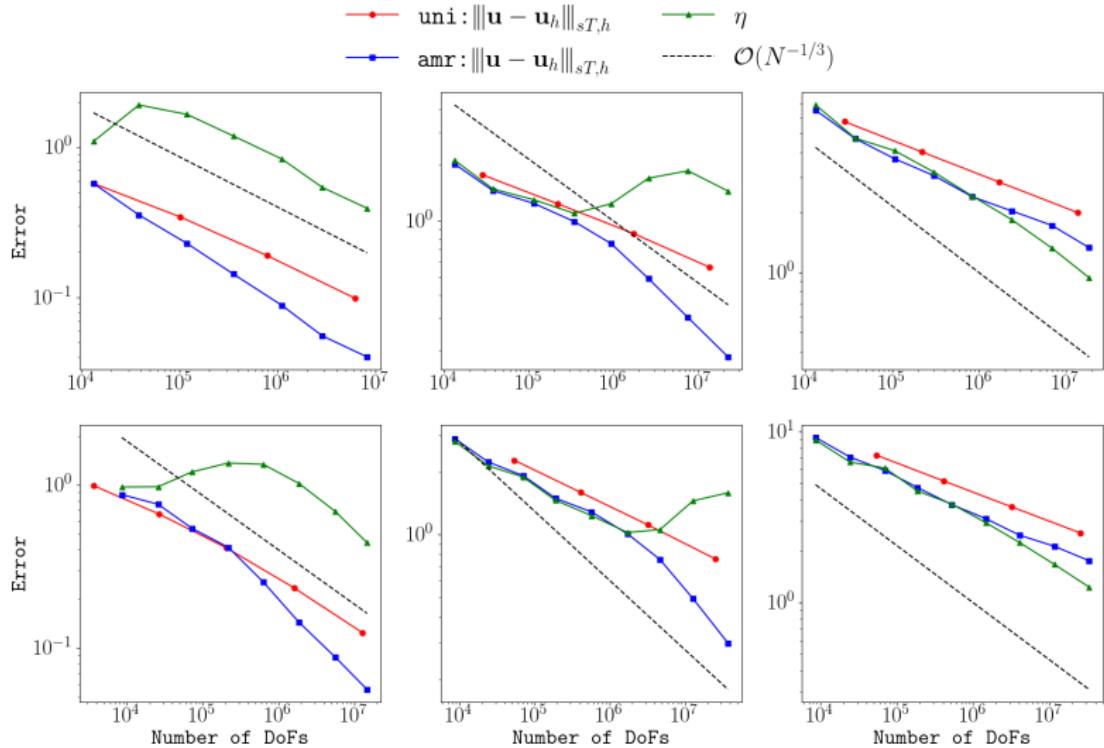
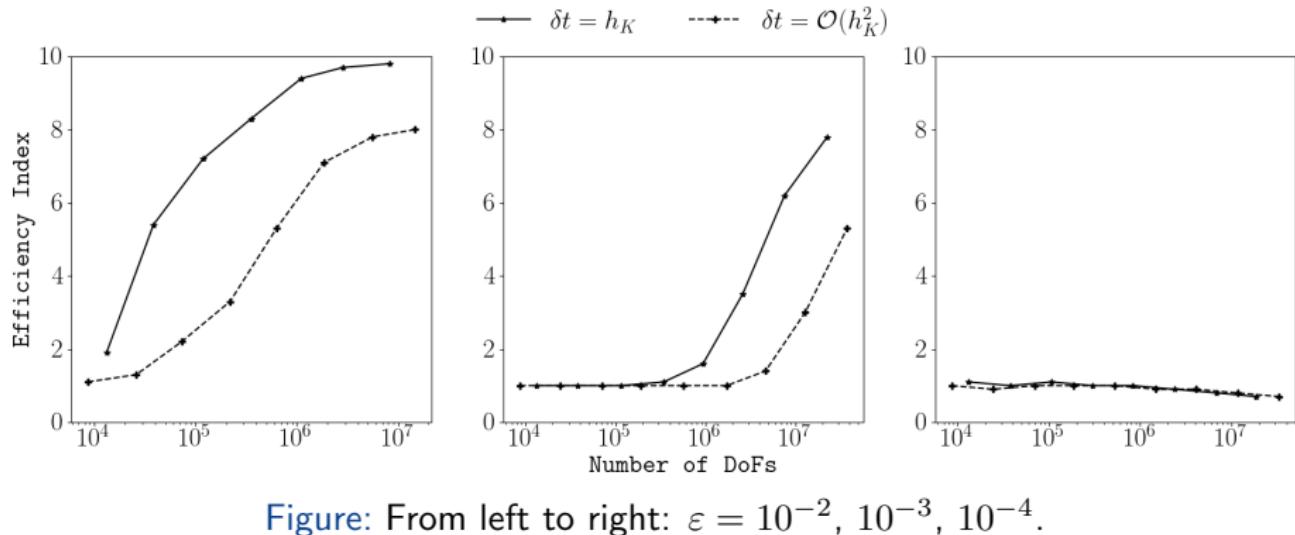


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An interior layer test case



Thank you!

Find codes, papers and more on my website:

gregw.xyz