

Adaptive Space-Time Isogeometric Analysis of Parabolic Evolution Problems

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- Globally stabilised space-time IgA schemes [Langer, Moore, and Neumüller, 2016]
- **Locally stabilized** space-time IgA schemes
- Adaptive space-time IgA schemes
- Numerical results
- Conclusions and roadmap

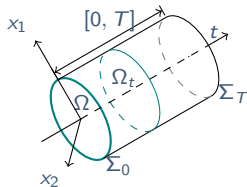
Intro: model problem and a posteriori error estimates and identity

Parabolic I-BVP model problem

Find $u : \bar{Q} \rightarrow \mathbb{R}$ satisfying the *linear parabolic initial-boundary value problem (I-BVP)*

$$\begin{aligned} \partial_t u - \operatorname{div}_x \nabla_x u &= f && \text{in } Q, \\ u(x, 0) &= u_0 && \text{on } \Sigma_0, \\ u &= u_D = 0 && \text{on } \Sigma, \end{aligned}$$

$$\begin{aligned} \Omega &\subset \mathbb{R}^d, d = \{1, 2, 3\}, T > 0 \\ Q &:= \Omega \times (0, T) \\ \partial Q &:= \Sigma \cup \bar{\Sigma}_0 \cup \bar{\Sigma}_T \\ \Sigma &:= \partial\Omega \times (0, T) \\ \Sigma_0 &:= \Omega \times \{0\} \\ \Sigma_T &:= \Omega \times \{T\} \end{aligned}$$



where ∂_t is the time derivative,
 $\Delta_x = \operatorname{div}_x \nabla_x$, div_x and ∇_x are
 Laplace, divergence, and gradient operators in space, resp.,
 $u_0 \in H_0^1(\Sigma_0)$ is a given initial state,
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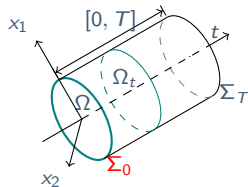
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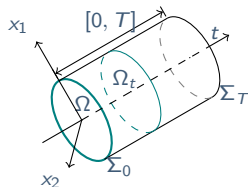
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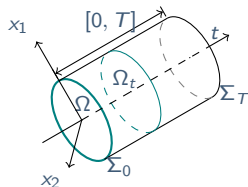
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Solvability results [Ladyzhenskaya, 1954]

Weak formulation:

Find $u \in H_0^{1,0}(Q) := \{v \in L^2(Q) : \nabla_x v \in [L^2(Q)]^d, v|_{\Sigma} = 0\}$ satisfying

$$(*) \quad a(u, w) = \ell(w), \quad \forall w \in H_{0,0}^{1,1}(Q) := \{v \in H_0^{1,0}(Q) : \partial_t v \in L^2(Q), v|_{\Sigma_T} = 0\},$$

where

$$a(u, w) := (\nabla_x u, \nabla_x w)_Q - (u, \partial_t w)_Q,$$

$$\ell(w) := (f, w)_Q + (u_0, w)_{\Sigma_0}.$$

If $f \in L^{2,1}(Q_T) := \{v \in L^1(Q) : \int_0^T \|v(\cdot, t)\|_{L^2(\Omega)} dt < \infty\}$ and $u_0 \in L^2(\Omega)$, then there \exists a unique weak solution $u \in H_0^{1,0}(Q)$ of (*) that also belongs to $V_0^{1,0} := C([0, T]; L^2(\Omega)) \cap H_0^{1,0}(Q)$.

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Functional a posteriori error analysis [Repin, 2002]

- For any $v \in H_0^{1,1}(Q)$, $y \in H^{\text{div}_x,0}(Q) := \{y \in [L^2(Q)]^{d+1} : \text{div}_x y \in L^2(Q)\}$, and $\beta > 0$, we have the following **functional a posteriori error estimate**:

$$\|u - v\|^2 := \|\nabla_x(u-v)\|_Q^2 + \|u - v\|_{\Sigma_T}^2 \leq \bar{M}^{1,2}(v, y; \beta)$$

with the majorant

$$\bar{M}^{1,2}(v, y; \beta) := (1+\beta) \underbrace{\|y - \nabla_x v\|_Q^2}_{\text{dual term } \bar{m}_d^1} + (1+\frac{1}{\beta}) C_{F\Omega}^2 \underbrace{\|f + \text{div}_x y - \partial_t v\|_Q^2}_{\text{equilibration/reliability term } \bar{m}_{\text{eq}}^1}.$$

- Main properties:
 - **universal** for any v from admissible functional space,
 - **computable**,
 - **reliable** and realistic w.r.t. the error, i.e., $1 \leq l_{\text{eff}} = \frac{\bar{M}}{\|u-v\|}$ is close to 1,
 - **efficient for adaptive strategies** $V_h \rightarrow V_{h_{\text{ref}}}$,
 - in the space-time setting, **allows fully-unstructured mesh adaptation**.

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Stronger solvability results [Ladyzhenskaya, 1954]

- If $f \in L^2(Q)$ and $u_0 \in H_0^1(\Omega)$, then the I-BVP is **uniquely solvable** in

$$H_0^{\Delta_x, 1}(Q) := \left\{ u \in H_0^{1,1}(Q) : \Delta_x u \in L^2(Q) \right\},$$

and u continuously depends on t in the $H_0^1(\Omega)$ -norm.

- Maximal parabolic regularity for $\partial_t u - \operatorname{div}_x(A(x, t)\nabla_x u) = f$:
for $f \in X = L^p((0, T); L^q(\Omega))$, $1 < p, q < \infty$ and $u_0 = 0$
there $\exists C > 0$, such that

$$\|\partial_t u\|_X + \|\operatorname{div}_x(A(x, t)\nabla_x u)\|_X \leq C \|f\|_X.$$

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Error identity [Anjam and Pauly, 2016]

For any $v \in H_0^{\Delta x, 1}(Q)$ approximating $u \in H_0^{\Delta x, 1}(Q)$, we have the **error identity**:

$$\begin{aligned} & \|\Delta_x(u - v)\|_Q^2 + \|\partial_t(u - v)\|_Q^2 + \|\nabla_x(u - v)\|_{\Sigma_T}^2 \\ & =: \|u - v\|_{\mathcal{L}, Q}^2 \equiv \mathbb{E}d^2(v) \\ & := \|\nabla_x(u_0 - v)\|_{\Sigma_0}^2 + \|\Delta_x v + f - \partial_t v\|_Q^2. \end{aligned}$$

Note:

- ⊕ reconstruction of $\mathbb{E}d^2(v)$ does not include time overhead
- ⊖ **extra regularity** $u, v \in H_0^{\Delta x, 1}(Q)$ is required (not practical for FEM) \Rightarrow but natural for IgA framework!

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Globally stabilized space-time IgA schemes

IgA framework [Hughes et al., 2005]

Physical domain

$Q \subset \mathbb{R}^{d+1}$ (single patch) is defined from

Parametric domain

$\hat{Q} := (0, 1)^{d+1}$ by the

Geometrical mapping

$\Phi : \hat{Q} \rightarrow Q = \Phi(\hat{Q}) \subset \mathbb{R}^{d+1}, \quad \Phi(\xi) = \sum_{i \in \mathcal{I}} \hat{B}_{i,p}(\xi) \mathbf{P}_i,$

- $\hat{B}_{i,p}, i \in \mathcal{I}$, are the B-splines, NURBS, THB-splines;
- $\{\mathbf{P}_i\}_{i \in \mathcal{I}} \in \mathbb{R}^{d+1}$ are the control points.

Set of facets:

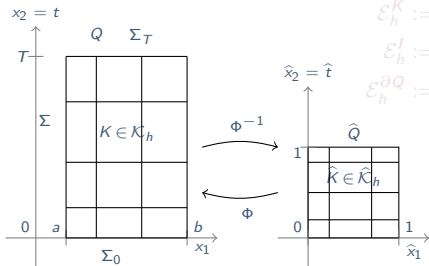
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where $\mathcal{E}_h^{\partial Q} := \mathcal{E}_h^\Sigma \cup \mathcal{E}_h^{\Sigma_T} \cup \mathcal{E}_h^{\Sigma_0}$, such that

$$\mathcal{E}_h^{\Sigma_*} := \{E : \exists K, K' \in \mathcal{K}_h : E = \partial K \cap \partial K', E \cap \Sigma_* \neq \emptyset\}.$$



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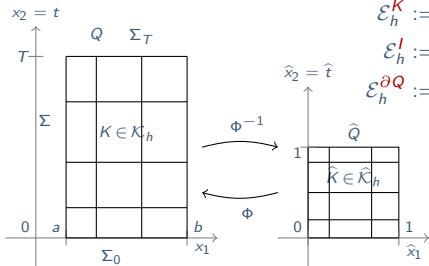
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Fundamentals [Bazilevs et. all., 2006],[Evans and Hughes, 2013]

Let $K \in \mathcal{K}_h$ and $h_K := \text{diam}_{K \in \mathcal{K}_h}(K)$, then the **inverse inequalities**

$$\|v_h\|_{\partial K} \leq C_{int,0} h_K^{-1/2} \|v_h\|_K \quad \text{and} \quad \|\nabla v_h\|_K \leq C_{int,1} h_K^{-1} \|v_h\|_K$$

hold for all $v_h \in V_h := \text{span} \{\phi_{h,i} := \widehat{B}_{i,p} \circ \Phi^{-1}\}_{i \in \mathcal{I}}$, where $C_{int,0}, C_{int,1} > 0$ are constants independent of K .

Let $K \in \mathcal{K}_h$, then the **scaled trace inequality**

$$\|v\|_{\partial K} \leq C_{tr} h_K^{-1/2} (\|v\|_K + h_K \|\nabla v\|_K)$$

hold for all $v \in H^1(K)$, where $C_{tr} > 0$ is a constant independent of K .

Approximation error estimates

Let $\ell, s \in \mathbb{N}$ be $0 \leq \ell \leq s \leq p + 1$, $u \in H_{0,0}^s(Q)$, and K and \underline{K} are element and its extension, resp. Then, $\exists \Pi_h : H_{0,0}^s(Q) \rightarrow V_{0h}$ such that

$$|v - \Pi_h v|_{H^\ell(K)}^2 \leq C_{\ell,s}^2 h_K^{2(s-\ell)} \sum_{i=0}^s c_K^{2(i-\ell)} |v|_{H^i(\underline{K})}^2, \quad \forall v \in L^2(Q) \cap H^\ell(\underline{K}),$$

where $c_K := \|\nabla_x \Phi\|_{L^\infty(\Phi^{-1}(\underline{K}))}$, and $C_{\ell,s} > 0$ is a constant dependent on s, ℓ, p , and the shape regularity of K , described by Φ and $\nabla_x \Phi$.

Stabilized variational identity for parabolic I-BVP

Testing the $-\Delta_x u + \partial_t u = f$ ($(d+1)$ -dimensional elliptic problem with convection in $(d+1)$ th direction) by the **upwind test function** with $\lambda, \mu \geq 0$

$$\lambda w + \mu \partial_t w, \quad w \in H_{0,0}^{\nabla_x \partial_t, 1}(Q) := \{w \in H_{0,0}^{\Delta_x, 1}(Q) : \nabla_x \partial_t w \in L^2(Q)\},$$

we obtain the variational identity

$$a(u, \lambda w + \mu \partial_t w) =: a_s(u, w) = \ell_s(w) := \ell(\lambda w + \mu \partial_t w)_Q, \quad \forall w \in H_{0,0}^{\nabla_x \partial_t, 1}(Q)$$

for the solution $u \in H_{0,0}^{\Delta_x, 1}(Q)$.

For any $v \in H_{0,0}^{\Delta_x, 1}$ approximating u , the error $u - v$ is measured in terms of the norm

$$\|u - v\|_s^2 := \lambda \left(\|\nabla_x(u - v)\|_Q^2 + \|u - v\|_{\Sigma_T}^2 \right) + \mu \left(\|\partial_t(u - v)\|_Q^2 + \|\nabla_x(u - v)\|_{\Sigma_T}^2 \right).$$

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Stabilized variational identity for parabolic I-BVP

Testing the $-\Delta_x u + \partial_t u = f$ ($(d+1)$ -dimensional elliptic problem with convection in $(d+1)$ th direction) by the **upwind test function** with $\lambda = 1$ and $\mu = \delta_h = \theta h$, $\theta > 0$, $h := \max_{K \in \mathcal{K}_h} \{h_K\}$

$$w + \delta_h \partial_t w, \quad w \in H_{0,0}^{\nabla_x \partial_t, 1}(Q) := \{w \in H_{0,0}^{\Delta_x, 1}(Q) : \nabla_x \partial_t w \in L^2(Q)\},$$

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$$\|u - v\|_{s,h}^2 := \|\nabla_x(u - v)\|_Q^2 + \|u - v\|_{\Sigma_T}^2 + \delta_h \left(\|\partial_t(u - v)\|_Q^2 + \|\nabla_x(u - v)\|_{\Sigma_T}^2 \right).$$

Stabilized space-time IgA scheme

IgA scheme [Langer, Moore, and Neumüller, 2016]

Find $u_h \in V_{0h} := \text{span} \{ \phi_{h,i} := \widehat{B}_{i,p} \circ \Phi^{-1} \}_{i \in \mathcal{I}} \cap H_0^1(Q) \subset H_{0,0}^{\Delta_x, 1}(Q)$, i.e. $p \geq 2$, satisfying

$$a_{s,h}(u_h, v_h) = \ell_{s,h}(v_h), \quad \forall v_h \in V_{0h},$$

where

$$a_{s,h}(u_h, v_h) := (\partial_t u_h, v_h + \delta_h \partial_t v_h)_Q + (\nabla_x u_h, \nabla_x (v_h + \delta_h \partial_t v_h))_Q,$$

$$\ell_{s,h}(v_h) := (f, v_h + \delta_h v_h)_Q,$$

Remark: For $u \in V_0^s, s \geq 2$ and $u_h \in V_{0h}$, there exists **a priori error estimates of the form**

$$\|u - u_h\|_{s,h} \leq C h^{r-1} \|u\|_{H^r(Q)}, \quad C > 0 \quad \text{and} \quad r = \min\{s, p + 1\}.$$

Majorant of the error norm $\| \cdot \|_s$

Theorem 1 [Langer, Matculevich, and Repin, 2016]

For any approximation $v \in H_0^{\Delta x, 1}(Q)$ to $u \in H_0^{\Delta x, 1}(Q)$ and for any $y \in H^{\text{div}_x, 0}(Q)$, the error $e = u - v$ can be estimated as follows:

$$\lambda (\|\nabla_x e\|_Q^2 + \|e\|_{\Sigma_T}^2) + \mu (\|\partial_t e\|_Q^2 + \|\nabla_x e\|_{\Sigma_T}^2) =: \|e\|_s^2$$

$$\leq \bar{M}_s^{1,2}(v, y; \beta, \alpha) := \lambda \bar{M}^{1,2}(v, y; \beta) + \mu \left((1 + \alpha) \|\text{div}_x r_d\|_Q^2 + \left(1 + \frac{1}{\alpha}\right) \|r_{\text{eq}}\|_Q^2 \right)$$

where

$$\begin{aligned} r_{\text{eq}}(v, y) &= f + \text{div}_x y - \partial_t v && \Leftarrow \partial_t u - \text{div}_x p = f, \\ r_d(v, y) &= y - \nabla_x v && \Leftarrow p = \nabla_x u, \\ \text{div}_x r_d(v, y) &= \text{div}_x y - \Delta_x v. \end{aligned}$$

$\lambda, \mu > 0$, and $\beta, \alpha > 0$ are auxiliary parameters.

Majorant of the error norm $\| \cdot \|_s$

Theorem 1 [Langer, Matculevich, and Repin, 2016]

For any approximation $v \in H_0^{\Delta x, 1}(Q)$ to $u \in H_0^{\Delta x, 1}(Q)$ and for any $\mathbf{y} \in H^{\text{div}_x, 0}(Q)$, the error $e = u - v$ can be estimated as follows:

$$\begin{aligned} & \lambda (\|\nabla_x e\|_Q^2 + \|e\|_{\Sigma_T}^2) + \mu (\|\partial_t e\|_Q^2 + \|\nabla_x e\|_{\Sigma_T}^2) =: \|e\|_s^2 \\ & \leq \overline{M}_s^{1,2}(v, \mathbf{y}; \beta, \alpha) := \lambda \overline{M}^{1,2}(v, \mathbf{y}; \beta) + \mu \left((1 + \alpha) \|\text{div}_x \mathbf{r}_d\|_Q^2 + \left(1 + \frac{1}{\alpha}\right) \|\mathbf{r}_{\text{eq}}\|_Q^2 \right) \end{aligned}$$

where

$$\begin{aligned} \mathbf{r}_{\text{eq}}(v, \mathbf{y}) &= \mathbf{f} + \text{div}_x \mathbf{y} - \partial_t v && \Leftarrow \partial_t u - \text{div}_x \mathbf{p} = \mathbf{f}, \\ \mathbf{r}_d(v, \mathbf{y}) &= \mathbf{y} - \nabla_x v && \Leftarrow \mathbf{p} = \nabla_x u, \\ \text{div}_x \mathbf{r}_d(v, \mathbf{y}) &= \text{div}_x \mathbf{y} - \Delta_x v. \end{aligned}$$

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$$\leq \bar{M}_s^{1,2}(v, \mathbf{y}; \beta, \alpha) := \lambda \bar{M}^{1,2}(v, \mathbf{y}; \beta) + \mu \left((1 + \alpha) \|\text{div}_x \mathbf{r}_d\|_Q^2 + (1 + \frac{1}{\alpha}) \|\mathbf{r}_{\text{eq}}\|_Q^2 \right)$$

where

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Majorant of the error norm $\| \cdot \|_{s,h}$

Theorem 1 [Langer, Matculevich, and Repin, 2016]

For any approximation $v \in H_0^{\Delta x,1}(Q)$ to $u \in H_0^{\Delta x,1}(Q)$ and for any $\mathbf{y} \in H^{\text{div}_x,0}(Q)$, the error $e = u - v$ can be estimated as follows:

$$\begin{aligned} (\|\nabla_x e\|_Q^2 + \|e\|_{\Sigma_T}^2) + \delta_h (\|\partial_t e\|_Q^2 + \|\nabla_x e\|_{\Sigma_T}^2) &=: \|e\|_{s,h}^2 \\ &\leq \overline{M}_s^{1,2}(v, \mathbf{y}; \beta, \alpha) := \overline{M}^{1,2}(v, \mathbf{y}; \beta) + \delta_h \left((1 + \alpha) \|\text{div}_x \mathbf{r}_d\|_Q^2 + \left(1 + \frac{1}{\alpha}\right) \|\mathbf{r}_{\text{eq}}\|_Q^2 \right) \end{aligned}$$

where

$$\begin{aligned} \mathbf{r}_{\text{eq}}(v, \mathbf{y}) &= f + \text{div}_x \mathbf{y} - \partial_t v && \Leftarrow \partial_t u - \text{div}_x \mathbf{p} = f, \\ \mathbf{r}_d(v, \mathbf{y}) &= \mathbf{y} - \nabla_x v && \Leftarrow \mathbf{p} = \nabla_x u, \\ \text{div}_x \mathbf{r}_d(v, \mathbf{y}) &= \text{div}_x \mathbf{y} - \Delta_x v. \end{aligned}$$

$\lambda, \mu > 0$, and $\beta, \alpha > 0$ are auxiliary parameters.

Locally stabilized space-time IgA schemes

Locally stabilized schemes

On each $K \in \mathcal{K}_h$, we test the PDE $\partial_t u - \Delta_x u = f$ with

$$v_h + \delta_K \partial_t v_h, \quad \delta_K = \theta_K h_K, \quad \text{where } \theta_K > 0 \quad \text{and} \quad h_K := \text{diam}(K),$$

yielding

$$(\partial_t u - \Delta_x u, v_h + \delta_K \partial_t v_h)_K = (f, v_h + \delta_K \partial_t v_h)_K, \quad \forall u \in H_0^{\Delta_x, 1}(Q), \quad \forall v_h \in V_{0h}.$$

Summing up $K \in \mathcal{K}_h$, we obtain

$$\begin{aligned} (\partial_t u - \Delta_x u, v_h)_Q + \sum_{K \in \mathcal{K}_h} \delta_K (\partial_t u - \Delta_x u, \partial_t v_h)_K &=: a_{loc}(u, v_h) \\ &= \ell_{loc}(v_h) := (f, v_h)_Q + \sum_{K \in \mathcal{K}_h} \delta_K (f, \partial_t v_h)_K. \end{aligned}$$

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$$(\partial_t u - \Delta_x u, v_h + \delta_K \partial_t v_h)_K = (f, v_h + \delta_K \partial_t v_h)_K, \quad \forall u \in H_0^{\Delta_x, 1}(Q), \quad \forall v_h \in V_{0h}.$$

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Locally Stabilized IgA counterpart

Find $u_h \in V_{0h}$ satisfying the variational IgA scheme

$$a_{loc,h}(u_h, v_h) = \ell_{loc,h}(v_h), \quad \forall u_h, v_h \in V_{0h},$$

where

$$\begin{aligned} a_{loc,h}(u_h, v_h) &:= (\partial_t u_h, v_h)_Q + (\nabla_x u_h, \nabla_x v_h)_Q \\ &+ \sum_{K \in \mathcal{K}_h} \delta_K \left((\partial_t u_h, \partial_t v_h)_K + (\nabla_x u_h, \nabla_x \partial_t v_h)_K \right) \\ &- \sum_{K \in \mathcal{K}_h} \delta_K \sum_{E \in \mathcal{E}_h^K \cap \mathcal{E}_h^I} (\mathbf{n}_x^E \cdot \nabla_x u_h, \partial_t v_h)_E. \end{aligned}$$

and

$$\ell_{loc,h}(v_h) := (f, v_h)_Q + \sum_{K \in \mathcal{K}_h} \delta_K (f, \partial_t v_h)_K.$$

V_{0h} -coercivity of $a_{loc,h}(\cdot, \cdot)$

Lemma (coercivity)

Let

$$\theta_K \in \left(0, \frac{h_K}{d C_{int,1}^2}\right], \quad K \in \mathcal{K}_h,$$

where $C_{int,1}$ is the constant in the 2nd inverse inequality.

Then, $a_{loc,h}(u_h, v_h) : V_{0h} \times V_{0h} \rightarrow \mathbb{R}$ is V_{0h} -coercive w.r.t. to the norm

$$\|v_h\|_{loc,h}^2 := \|\nabla_x v_h\|_Q^2 + \frac{1}{2} \|v_h\|_{\Sigma_T}^2 + \sum_{K \in \mathcal{K}_h} \delta_K \|\partial_t v_h\|_K^2,$$

i.e., there exists a constant $\mu_{c,loc} > 0$ independent on K such that

$$a_{loc,h}(u_h, v_h) \geq \mu_{c,loc} \|v_h\|_{loc,h}^2.$$

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Uniform boundedness of $a_{loc,h}(\cdot, \cdot)$ on $V_{0h,*} \times V_{0h}$

Let $V_{0h,*} = H_0^{\Delta x, 1}(Q) + V_{0h}$ equipped with the norm

$$\|v\|_{loc,h,*}^2 := \|v\|_{loc,h}^2 + \sum_{K \in \mathcal{K}_h} (\delta_K^{-1} \|v\|_K^2 + \delta_K \|\Delta_x v\|_K^2).$$

Lemma (boundedness)

Let $\theta_K \in \left(0, \frac{h_K}{d C_{int,1}^2}\right]$, $K \in \mathcal{K}_h$. Then, $a_{loc,h}(\cdot, \cdot)$ is **uniformly bounded** on $V_{0h,*} \times V_{0h}$, i.e., there exist a constant $\mu_{b,loc} > 0$ independent on h_K such that

$$|a_{loc,h}(v, v_h)| \leq \mu_{b,loc} \|v\|_{loc,h,*} \|v_h\|_{loc,h}, \quad \forall v \in V_{0h,*}, \quad \forall v_h \in V_{0h}.$$

Approximation properties and consistency

Lemma (approximation error estimates)

Let $l, s \in \mathbb{N}$ be $1 \leq l \leq s \leq p + 1$, and $u \in H_{0,0}^s(Q)$. Then, $\exists \Pi_h : H_{0,0}^s(Q) \rightarrow V_{0h}$ and $C_1, C_2 > 0$, such that **a priori error estimates** hold

$$\|u - \Pi_h u\|_{loc,h}^2 \leq C_1 \sum_{K \in \mathcal{K}_h} h_K^{2(s-1)} \sum_{i=0}^s c_K^{2i} |u|_{H^i(\underline{K})}^2,$$

$$\|u - \Pi_h u\|_{loc,h,*}^2 \leq C_2 \sum_{K \in \mathcal{K}_h} h_K^{2(s-1)} \sum_{i=0}^s c_K^{2i} |u|_{H^i(\underline{K})}^2.$$

where $K \in \mathcal{K}_h$ is the mesh element and \underline{K} is its support extension on the physical domain.

Lemma (consistency)

Let $p \geq 2$. If the solution $u \in H_0^{\Delta x, 1}(Q)$, then it satisfies the **consistency variational identity**

$$a_{loc,h}(u, v_h) = \ell_{loc,h}(v_h), \quad \forall v_h \in V_{0h}.$$

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A priori error estimate

Theorem (a priori error estimates)

Let $p \geq 2$, $u \in H_0^{\Delta x, 1}(Q) \cap H_0^s$, $s \geq 2$, be the exact solution, and $u_h \in V_{0h}$ be a solution of discrete IgA scheme

$$a_{loc,h}(u_h, v_h) = \ell_{loc,h}(v_h), \quad \forall u_h, v_h \in V_{0h} \quad \text{with} \quad \theta_K \in \left(0, \frac{h_K}{d C_{int,1}^2}\right], \quad K \in \mathcal{K}_h.$$

Then, the **a priori error estimate**

$$\|u - u_h\|_{loc,h}^2 \leq C \sum_{K \in \mathcal{K}_h} h_K^{2(r-1)} \sum_{i=0}^r c_K^{2i} |u|_{H^i(K)}^2, \quad r = \min\{s, p + 1\},$$

holds, where p denotes the polynomial degree of the THB-splines,

$C = \left(1 + \frac{\mu_{loc,b}}{\mu_{loc,c}}\right) C_2$ is a constant independent of h_K ,

$K \in \mathcal{K}_h$ and its support extension \underline{K} , and

$\mu_{loc,b}$ and $\mu_{loc,c}$ are constant in boundedness and coercivity inequalities, respectively.

Adaptive space-time IgA schemes

Majorants for the heat equation with Dirichlet BC

For given $f \in L^2(Q)$ and $u_0 \in H_0^1(\Omega)$, find $u \in H_0^{\Delta x, 1}(Q)$

$$u_t - \Delta_x u = f \text{ in } Q, \quad u = u_D \text{ on } \Sigma, \quad u(0, x) = u_0 \text{ on } \Sigma_0.$$

The error $e = u - v = u - u_h$ is tracked by the norms

$$\|e\|_{loc, h}^2 := \|\nabla_x e\|_Q^2 + \frac{1}{2} \|e\|_{\Sigma_T}^2 + \sum_{K \in \mathcal{K}_h} \delta_K \|\partial_t e\|_K^2, \quad \delta_K = \theta_K h_K, \quad \theta_K \in \left(0, \frac{h_K}{d C_{int, 1}^2}\right].$$

For any $v \in H_0^{\Delta x, 1}(Q)$ and $y \in H(Q, \text{div}_x)$, $w \in H_0^{\Delta x, 1}(Q)$, and $\beta, \alpha > 0$, we have a **posteriori estimates**

$$\|e\|^2 := \|\nabla_x e\|_Q^2 + \|e\|_{\Sigma_T}^2 \leq \bar{M}^{1,2}(v, y; \beta) \quad (\bar{M}^{\mathbb{H}, 2}(v, y, w; \beta^{\mathbb{H}}))$$

and **error identity**

$$\|e\|_{\mathcal{L}, Q}^2 := \|\Delta_x e\|_Q^2 + \|\partial_t e\|_Q^2 + \|\nabla_x e\|_{\Sigma_T}^2 \equiv \mathbb{H}^2(v).$$

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$$\|e\|^2 := \|\nabla_x e\|_Q^2 + \|e\|_{\Sigma_T}^2 \leq \bar{M}^{I, 2}(v, y; \beta) \quad (\bar{M}^{II, 2}(v, y, w; \beta^{II}))$$

and **error identity**

$$\|e\|_{\mathcal{L}, Q}^2 := \|\Delta_x e\|_Q^2 + \|\partial_t e\|_Q^2 + \|\nabla_x e\|_{\Sigma_T}^2 \equiv \mathbb{E}d^2(v).$$

Reconstruction of optimal $\overline{M}^I(v, \mathbf{y}; \beta^I)$

Solving $\{\mathbf{y}_{\min}, \beta_{\min}^I\} := \arg \inf_{\beta^I > 0} \inf_{\mathbf{y} \in H(Q, \text{div}_x)} \overline{M}^{I,2}(v, \mathbf{y}; \beta^I)$, where

$$\overline{M}^{I,2}(v, \mathbf{y}; \beta^I) := (1 + \beta^I) \underbrace{\|\mathbf{y} - \nabla_x v\|_Q^2}_{\overline{m}_d^{I,2}} + \left(1 + \frac{1}{\beta^I}\right) C_{F\Omega}^2 \underbrace{\|f + \text{div}_x \mathbf{y} - \partial_t v\|_Q^2}_{\overline{m}_{\text{eq}}^{I,2}}$$

leads to

- the auxiliary variation problem for the optimal \mathbf{y}_{\min} , i.e.,

$$\frac{C_{F\Omega}^2}{\beta_{\min}^I} (\text{div}_x \mathbf{y}_{\min}, \text{div}_x \eta)_Q + (\mathbf{y}_{\min}, \eta)_Q = - \frac{C_{F\Omega}^2}{\beta_{\min}^I} (f - \partial_t v, \text{div}_x \eta)_Q + (\nabla_x v, \eta)_Q,$$

- with the optimal $\beta_{\min}^I := \frac{C_{F\Omega} \overline{m}_{\text{eq}}^I}{\overline{m}_d^I}$.

IgA spaces for u_h approximation

$$\widehat{V}_h \equiv \widehat{S}_h^p := \text{span} \{ \widehat{B}_{i,p} \},$$

$$u_h \in V_h \equiv S_h^p := \{ \widehat{V}_h \circ \Phi^{-1} \} \cap H_{UD}^1(Q) := \text{span} \{ \phi_{h,i} := \widehat{B}_{i,p} \circ \Phi^{-1} \}_{i \in \mathcal{I}} \cap H_{UD}^1(Q).$$

Generated approximation u_h is presented as

$$u_h(x) = \sum_{i \in \mathcal{I}} \underline{u}_i \phi_{h,i}(x), \quad \underline{u}_h := [u_i]_{i \in \mathcal{I}} \in \mathbb{R}^{|\mathcal{I}|},$$

where \underline{u}_h is a vector of DOFs defined by a system

$$K_h \underline{u}_h = f_h, \quad t_{\text{as}}(u_h) + t_{\text{sol}}(u_h)$$

$$K_h := [a_{s,h}(\phi_{h,i}, \phi_{h,j})]_{i,j}^{\mathcal{I}},$$

$$f_h := [\ell_{s,h}(\phi_{h,i})]_i^{\mathcal{I}}.$$

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where \underline{u}_h is a vector of DOFs defined by a system

$$\mathbf{K}_h \underline{u}_h = \mathbf{f}_h, \quad : t_{\text{as}}(u_h) + t_{\text{sol}}(u_h)$$

$$\mathbf{K}_h := [a_{s,h}(\phi_{h,i}, \phi_{h,j})]_{i,j}^{\mathcal{I}},$$

$$\mathbf{f}_h := [\ell_{s,h}(\phi_{h,i})]_i^{\mathcal{I}}.$$

IgA spaces for the flux reconstruction y_h

$$\hat{Y}_h \equiv \oplus^{d+1} \hat{S}_h^q,$$

$$\mathbf{y}_h = \begin{bmatrix} y_h^{(1)} \\ \dots \\ y_h^{(d+1)} \end{bmatrix} \in \mathbf{Y}_h \equiv \oplus^{d+1} \mathcal{S}_h^q := \{ \hat{Y}_h \circ \Phi^{-1} \} = \text{span} \{ \psi_{h,i} := [\hat{B}_{i,q}]^{d+1} \circ \Phi^{-1} \}_{i \in \mathcal{I}}$$

Generated reconstruction of y_h is presented as

$$y_h(x) := \sum_{i \in \mathcal{I} \times (d+1)} \underline{y}_{h,i} \psi_{h,i}(x),$$

where $\underline{y}_h := [\underline{y}_{h,i}]_{i \in \mathcal{I} \times (d+1)} \in \mathbb{R}^{(d+1)|\mathcal{I}|}$ is a vector of DOFs of y_h defined by a system

$$(C_{F\Omega}^2 \text{Div}_h + \beta M_h) \underline{y}_h = -C_{F\Omega}^2 z_h + \beta g_h, \quad : t_{\text{as}}(y_h) + t_{\text{sol}}(y_h)$$

with

$$\text{Div}_h := [(\text{div}_x \psi_i, \text{div}_x \psi_j)]_{i,j=1}^{(d+1)|\mathcal{I}|}, \quad z_h := [(f - \partial_t v, \text{div}_x \psi_j)]_{j=1}^{(d+1)|\mathcal{I}|},$$

$$M_h := [(\psi_i, \psi_j)]_{i,j=1}^{(d+1)|\mathcal{I}|}, \quad g_h := [(\nabla_x v, \psi_j)]_{j=1}^{(d+1)|\mathcal{I}|}.$$

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$$\mathbf{y}_h = \begin{bmatrix} y_h^{(1)} \\ \dots \\ y_h^{(d+1)} \end{bmatrix} \in Y_h \equiv \oplus^{d+1} S_h^q := \{ \hat{Y}_h \circ \Phi^{-1} \} = \text{span} \{ \psi_{h,i} := [\hat{B}_{i,q}]^{d+1} \circ \Phi^{-1} \}_{i \in \mathcal{I}}$$

Generated reconstruction of y_h is presented as

$$\mathbf{y}_h(x) := \sum_{i \in \mathcal{I} \times (d+1)} \underline{y}_{h,i} \psi_{h,i}(x),$$

where $\underline{y}_h := [\underline{y}_{h,i}]_{i \in \mathcal{I} \times (d+1)} \in \mathbb{R}^{(d+1)|\mathcal{I}|}$ is a vector of DOFs of y_h defined by a system

$$(\mathcal{C}_{\Gamma\Omega}^2 \text{Div}_h + \beta M_h) \underline{y}_h = -\mathcal{C}_{\Gamma\Omega}^2 z_h + \beta g_h, \quad : \mathbf{t}_{\text{as}}(y_h) + \mathbf{t}_{\text{sol}}(y_h)$$

with

$$\text{Div}_h := [(\text{div}_x \psi_i, \text{div}_x \psi_j)]_{i,j=1}^{(d+1)|\mathcal{I}|}, \quad z_h := [(f - \partial_t v, \text{div}_x \psi_j)]_{j=1}^{(d+1)|\mathcal{I}|},$$

$$M_h := [(\psi_i, \psi_j)]_{i,j=1}^{(d+1)|\mathcal{I}|}, \quad g_h := [(\nabla_x v, \psi_j)]_{j=1}^{(d+1)|\mathcal{I}|}.$$

Single refinement step for u_h approximation

Input: \mathcal{K}_h {discretization of Q }, $\text{span} \{ \phi_{h,i} \}, i = 1, \dots, |\mathcal{I}|$ { V_h -basis}

APPROXIMATE:

■ compute u_h : ASSEMBLE and SOLVE $K_h \underline{u}_h = \underline{f}_h$: $t_{as}(u_h) + t_{sol}(u_h)$

Evaluate $e = u - u_h$ in terms of $\|e\|$, $\|e\|_{loc,h}$, and $\|e\|_{\mathcal{L}}$

ESTIMATE:

■ compute $\bar{M}^I(u_h, y_h)$: $t_{as}(y_h) + t_{sol}(y_h)$

■ compute $\bar{M}^{II}(u_h, y_h, w_h)$: $t_{as}(w_h) + t_{sol}(w_h)$

■ compute $\mathbb{E}d(u_h)$

MARK: Using marking $\mathbb{M}_{BULK}(\sigma)$, select elements K of mesh \mathcal{K}_h that must be refined

REFINE: Execute the refinement strategy $\mathcal{K}_{h_{ref}} = R(\mathcal{K}_h)$

Output: $\mathcal{K}_{h_{ref}}$ {refined discretization of Q }

Choice of B-Splines (THB-Splines) for y_h and w_h

We use the idea from [Kleiss, Tomar, 2015]:

- $u_h \in V_h \equiv S_h^p$

- u_h is approx. on \mathcal{K}_h

- $y_h \in Y_{Mh} \equiv \oplus^{d+1} S_{Mh}^q$

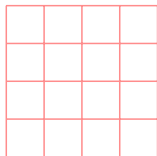
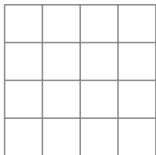
- $q \gg p,$

- y_{Mh} is reconstructed on $\mathcal{K}_{Mh}, M \in \mathbb{N}^+$

- $w_h \in W_{Lh} \equiv S_{Lh}^r$

- $r \gg p,$

- w_{Lh} is reconstructed on $\mathcal{K}_{Lh}, L \in \mathbb{N}^+$



Choice of B-Splines (THB-Splines) for y_h and w_h

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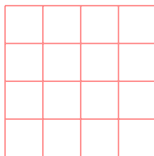
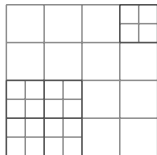
- $q \gg p,$

- $r \gg p,$

- u_h is approx. on \mathcal{K}_h

- y_{Mh} is reconstructed on $\mathcal{K}_{Mh}, M \in \mathbb{N}^+$

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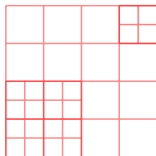
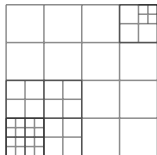
- $q \gg p,$

- $r \gg p,$

- u_h is approx. on \mathcal{K}_h

- y_{Mh} is reconstructed on $\mathcal{K}_{Mh}, M \in \mathbb{N}^+$

- w_{Lh} is reconstructed on $\mathcal{K}_{Lh}, L \in \mathbb{N}^+$



Choice of B-Splines (THB-Splines) for y_h and w_h

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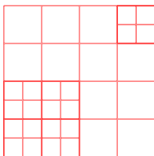
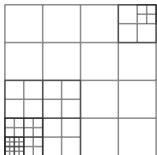
- $q \gg p,$

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- $w_h \in W_{Lh} \equiv S_{Lh}^r$

- $r \gg p,$

- w_{Lh} is reconstructed on $\mathcal{K}_{Lh}, L \in \mathbb{N}^+$



Numerical results

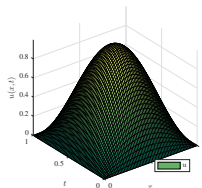
Example 1

Given data of 1d+t dimensional problem:

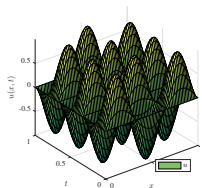
- $\Omega = (0, 1), T = 1$
- $u = \sin(k_1 \pi x) \sin(k_2 \pi t)$
- $f = \sin(k_1 \pi x) (k_2 \pi \cos(k_2 \pi t) + k_1^2 \pi^2 \sin(k_2 \pi t))$
- $u_D = 0$

Discretization:

- $u_h \in S_h^2$ and $u_h \in S_h^3$
- **Example 1-1:** $k_1 = k_2 = 1$:
- **Example 1-2:** $k_1 = 3, k_2 = 6$:



$k_1 = k_2 = 1$



$k_1 = 3, k_2 = 6$

Example 1-1. Adaptive refinement for $u_h \in S_h^2$ and $u_h \in S_h^3$

# ref.	$\ e\ _Q$	$I_{\text{eff}}(\overline{M}^I)$	$I_{\text{eff}}(\overline{M}^{II})$	$\ e\ _{loc,h}$	$\ e\ _{\mathcal{L}}$	$I_{\text{eff}}(\mathbb{E}d)$	e.o.c. ($\ e\ _{loc,h}$)	e.o.c. ($\ e\ _{\mathcal{L}}$)
(a) $u_h \in S_h^2, y_h \in \oplus^2 S_{7h}^4, \text{ and } w_h \in S_{7h}^4$								
2	2.9034e-03	1.94	1.17	3.0649e-03	2.9197e-01	1.00	2.38	1.40
4	3.3878e-04	3.14	1.33	3.5057e-04	9.3154e-02	1.00	1.96	1.07
8	9.2649e-06	5.78	3.23	9.2835e-06	1.7351e-02	1.00	3.79	1.79

(b) $u_h \in S_h^3, y_h \in \oplus^2 S_{5h}^6, \text{ and } w_h \in S_{5h}^6$								
2	4.9924e-03	1.31	1.04	5.0700e-03	1.1918e-01	1.00	5.08	4.18
4	1.3562e-04	1.64	1.30	1.3591e-04	8.9725e-03	1.00	3.56	2.89
8	3.5507e-07	3.44	1.24	3.5535e-07	1.6376e-04	1.00	3.11	2.13

Efficiency of $\overline{M}^I, \overline{M}^{II}$, and $\mathbb{E}d$ for $\sigma = 0.4$ ($N_{\text{ref},0} = 3$).

Example 1-1. Adaptive refinement for $u_h \in S_h^2$ and $u_h \in S_h^3$

# ref.	d.o.f.			t_{as}			t_{sol}			$\frac{t_{appr.}}{t_{er.est.}}$
	u_h	y_h	w_h	u_h	y_h	w_h	u_h	y_h	w_h	
(a) $u_h \in S_h^2, y_h \in \oplus^2 S_{7h}^4$, and $w_h \in S_{7h}^4$										
6	12935	288	144	1.55e+01	3.97e-01	3.83e-01	2.17e+00	2.30e-03	1.37e-03	44.25
7	34037	288	144	4.90e+01	3.98e-01	3.73e-01	9.58e+00	3.36e-03	1.42e-03	145.95
8	61258	288	144	9.37e+01	3.80e-01	3.62e-01	2.42e+01	2.10e-03	1.83e-03	308.55
				$t_{as}(u_h)$ 258.63	$t_{as}(y_h)$ 1.05	$t_{as}(w_h)$ 1.00	$t_{sol}(u_h)$ 13252.51	$t_{sol}(y_h)$ 1.15	$t_{sol}(w_h)$ 1.00	
(b) $u_h \in S_h^3, y_h \in \oplus^2 S_{5h}^5$, and $w_h \in S_{5h}^5$										
6	13742	338	169	1.62e+01	7.03e-01	7.03e-01	2.11e+00	2.53e-03	1.43e-03	25.95
7	35091	644	322	5.36e+01	5.65e+00	5.52e+00	1.10e+01	9.31e-03	5.29e-03	11.41
8	78561	744	372	1.91e+02	5.61e+00	5.03e+00	2.40e+01	2.51e-02	7.56e-03	38.15
				$t_{as}(u_h)$ 37.97	$t_{as}(y_h)$ 1.11	$t_{as}(w_h)$ 1.00	$t_{sol}(u_h)$ 3168.34	$t_{sol}(y_h)$ 3.31	$t_{sol}(w_h)$ 1.00	

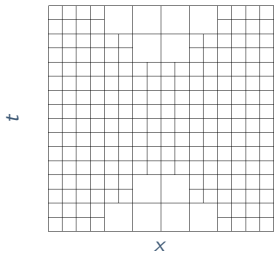
Assembling and solving time spent for the systems defining d.o.f. of u_h, y_h , and w_h .

Example 1-1. Comparison of meshes for $\mathbb{M}_{\text{BULK}}(0.4)$

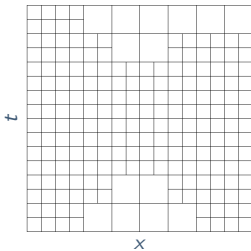
ref. based on
true error $\|u - u_h\|_{loc,h,K}^2$

ref. based on
true error $\|u - u_h\|_K^2$

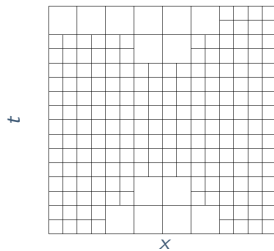
ref. based on
indicator $\|y_h - \nabla_x u_h\|_K^2$



ref. 1



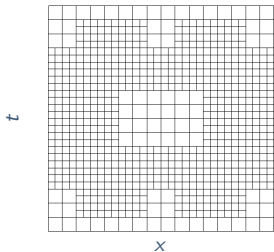
ref. 1



ref. 1

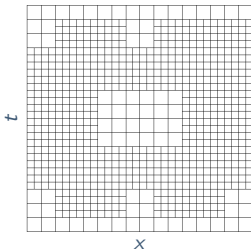
Example 1-1. Comparison of meshes for $\mathbb{M}_{\text{BULK}}(0.4)$

ref. based on
true error $\|u - u_h\|_{loc,h,K}^2$



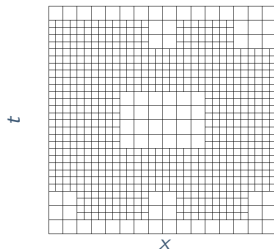
ref. 2

ref. based on
true error $\|u - u_h\|_K^2$



ref. 2

ref. based on
indicator $\|y_h - \nabla_x u_h\|_K^2$



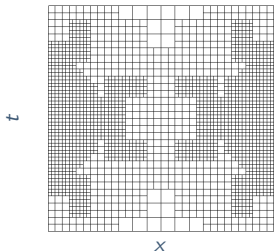
ref. 2

Example 1-1. Comparison of meshes for $\mathbb{M}_{\text{BULK}}(0.4)$

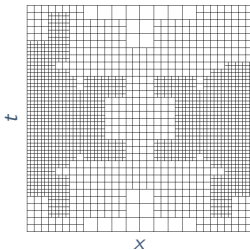
ref. based on
true error $\|u - u_h\|_{loc,h,K}^2$

ref. based on
true error $\|u - u_h\|_K^2$

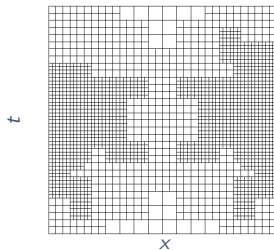
ref. based on
indicator $\|y_h - \nabla_x u_h\|_K^2$



ref. 3



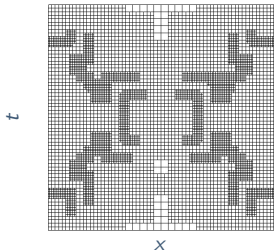
ref. 3



ref. 3

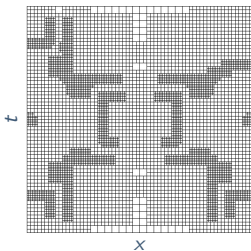
Example 1-1. Comparison of meshes for $\mathbb{M}_{BULK}(0.4)$

ref. based on
true error $\|u - u_h\|_{loc,h,K}^2$



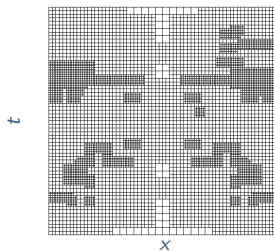
ref. 4

ref. based on
true error $\|u - u_h\|_K^2$



ref. 4

ref. based on
indicator $\|y_h - \nabla_x u_h\|_K^2$



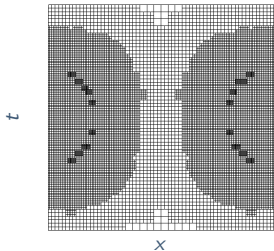
ref. 4

Example 1-1. Comparison of meshes for $\mathbb{M}_{BULK}(0.4)$

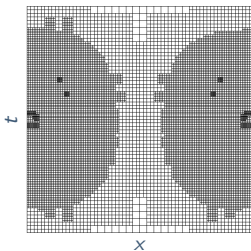
ref. based on
true error $\|u - u_h\|_{loc,h,K}^2$

ref. based on
true error $\|u - u_h\|_K^2$

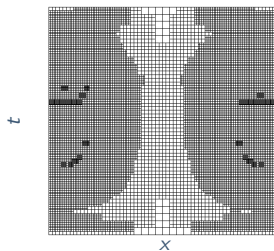
ref. based on
indicator $\|y_h - \nabla_x u_h\|_K^2$



ref. 5



ref. 5



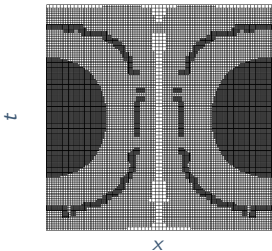
ref. 5

Example 1-1. Comparison of meshes for $\mathbb{M}_{BULK}(0.4)$

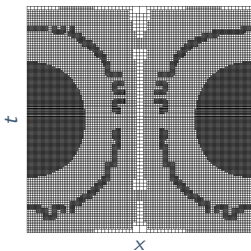
ref. based on
true error $\|u - u_h\|_{loc,h,K}^2$

ref. based on
true error $\|u - u_h\|_K^2$

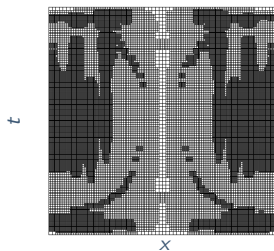
ref. based on
indicator $\|y_h - \nabla_x u_h\|_K^2$



ref. 6

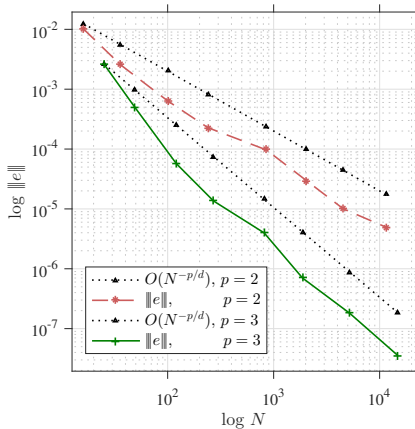


ref. 6



ref. 6

Example 1-1. Error order of convergence



The e.o.c. for $k_1 = k_2 = 1$.

Example 1-2. Adaptive refinement, $u_h \in S_h^2$, $\mathbf{y}_h \in \oplus^2 S_{5h}^7$, and $w_h \in S_{5h}^7$

# ref.	$\ e\ _Q$	$I_{\text{eff}}(\overline{M}^I)$	$I_{\text{eff}}(\overline{M}^{II})$	$\ e\ _{loc,h}$	$\ e\ _{\mathcal{L}}$	$I_{\text{eff}}(\mathbb{E}d)$	e.o.c. ($\ e\ _{loc,h}$)	e.o.c. ($\ e\ _{\mathcal{L}}$)
(a) $M_{\text{BULK}}(0.4)$								
2	5.7161e-01	2.11	1.38	5.7163e-01	6.2371e+01	1.00	2.99	1.19
3	1.3927e-01	5.77	2.20	1.3928e-01	3.1026e+01	1.00	2.30	1.14
8	1.2298e-03	1.44	1.16	1.2298e-03	2.6917e+00	1.00	5.60	2.30
(b) $M_{\text{BULK}}(0.6)$								
2	5.7161e-01	2.11	1.38	5.7163e-01	6.2371e+01	1.00	2.99	1.19
3	1.7942e-01	4.69	1.96	1.7945e-01	3.2971e+01	1.00	2.18	1.20
8	2.7492e-03	1.44	1.15	2.7492e-03	4.0721e+00	1.00	4.75	1.91

Efficiency of \overline{M}^I , \overline{M}^{II} , and $\mathbb{E}d$ for $M_{\text{BULK}}(0.4)$ and $M_{\text{BULK}}(0.6)$ ($N_{\text{ref},0} = 3$).

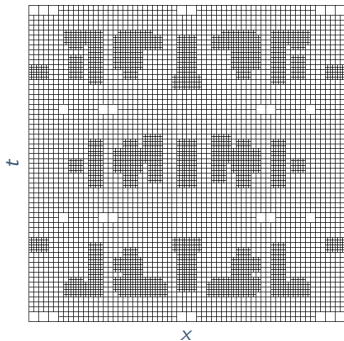
Example 1-2. Adaptive refinement, $u_h \in S_h^2$, $y_h \in \oplus^2 S_{5h}^7$, and $w_h \in S_{5h}^7$

# ref.	d.o.f.			t_{as}			t_{sol}			$\frac{t_{appr.}}{t_{er.est.}}$
	u_h	y_h	w_h	u_h	y_h	w_h	u_h	y_h	w_h	
(a) $M_{BULK}(0.4)$										
6	30101	450	225	5.99e+01	2.29e+00	2.92e+00	3.57e+00	8.52e-03	4.33e-03	12.14
7	86849	1058	529	3.57e+02	9.30e+00	9.41e+00	1.11e+01	5.19e-02	3.47e-02	19.58
8	141987	2850	1425	6.36e+02	6.50e+01	5.91e+01	2.56e+01	3.00e-01	1.29e-01	5.31
				$t_{as}(u_h) : t_{as}(y_h) : t_{as}(w_h)$			$t_{sol}(u_h) : t_{sol}(y_h) : t_{sol}(w_h)$			
				10.76	1.10	1.00	198.84	2.32	1.00	
(b) $M_{BULK}(0.6)$										
6	15436	450	225	2.61e+01	2.36e+00	2.41e+00	1.77e+00	1.45e-02	3.12e-03	11.73
7	35745	1058	529	8.99e+01	9.86e+00	1.01e+01	4.68e+00	7.06e-02	4.12e-02	9.52
8	52453	2498	1249	1.05e+02	8.03e+01	7.08e+01	7.38e+00	3.47e-01	1.66e-01	1.39
				$t_{as}(u_h) : t_{as}(y_h) : t_{as}(w_h)$			$t_{sol}(u_h) : t_{sol}(y_h) : t_{sol}(w_h)$			
				1.49	1.13	1.00	44.46	2.09	1.00	

Assembling and solving time spent for the systems defining d.o.f. of u_h , y_h , and w_h .

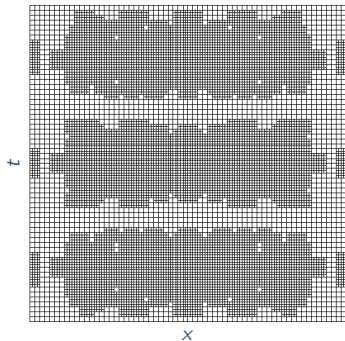
Example 1-2. Comparison of meshes for $M_{BULK}(0.6)$ and $M_{BULK}(0.4)$

mesh obtained with criterion $M_{BULK}(0.6)$



ref. 4

mesh obtained with criterion $M_{BULK}(0.4)$

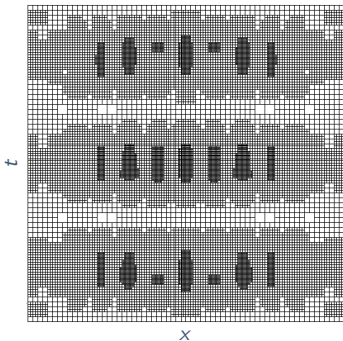


ref. 4

$$u_h \in S_h^2, \mathbf{y}_h \in \oplus^2 S_{5h}^7, \text{ and } w_h \in S_{5h}^7.$$

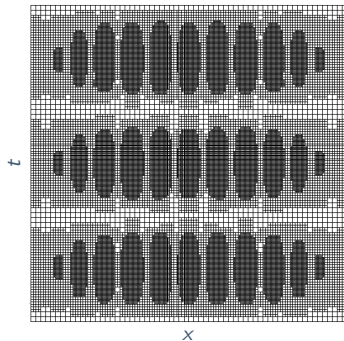
Example 1-2. Comparison of meshes for $M_{BULK}(0.6)$ and $M_{BULK}(0.4)$

mesh obtained with criterion $M_{BULK}(0.6)$



ref. 5

mesh obtained with criterion $M_{BULK}(0.4)$

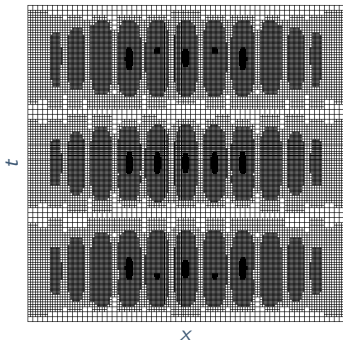


ref. 5

$$u_h \in S_h^2, \mathbf{y}_h \in \oplus^2 S_{5h}^7, \text{ and } w_h \in S_{5h}^7.$$

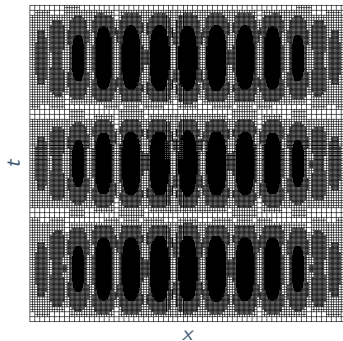
Example 1-2. Comparison of meshes for $M_{BULK}(0.6)$ and $M_{BULK}(0.4)$

mesh obtained with criterion $M_{BULK}(0.6)$



ref. 6

mesh obtained with criterion $M_{BULK}(0.4)$

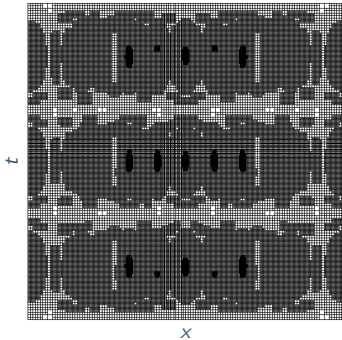


ref. 6

$$u_h \in S_h^2, \mathbf{y}_h \in \bigoplus^2 S_{5h}^7, \text{ and } w_h \in S_{5h}^7.$$

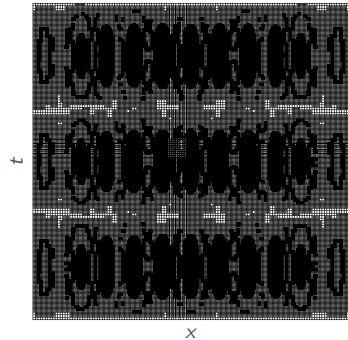
Example 1-2. Comparison of meshes for $M_{BULK}(0.6)$ and $M_{BULK}(0.4)$

mesh obtained with criterion $M_{BULK}(0.6)$



ref. 7

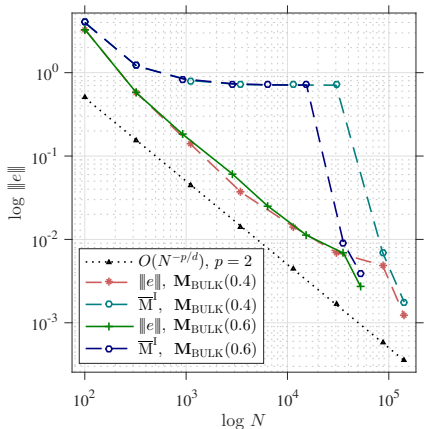
mesh obtained with criterion $M_{BULK}(0.4)$



ref. 7

$$u_h \in S_h^2, \mathbf{y}_h \in \oplus^2 S_{5h}^7, \text{ and } w_h \in S_{5h}^7.$$

Example 1-2. Error order of convergence



The e.o.c. for $k_1 = 6, k_2 = 3$.

Example 2. Moving spatial domains

- $Q := \{(x, t) \in \mathbb{R}^{d+1} : x \in \Omega(t), t \in (0, T)\}$, where

$$\Omega(t) = \{x \in \mathbb{R}^d : a(t) < x < b(t)\}, t \in (0, T)$$

$$a(t) = 0.5 t(1 - t),$$

$$b(t) = 1 - a(t), \text{ and}$$

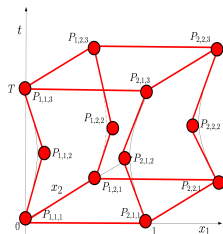
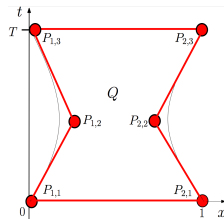
$$T = 1$$

1d+t:

- $$u(x, t) = \sin(\pi x) \sin(\pi t),$$
- $$f(x, t) = \pi \sin(\pi x) (\cos(\pi t) + \pi \sin(\pi t))$$

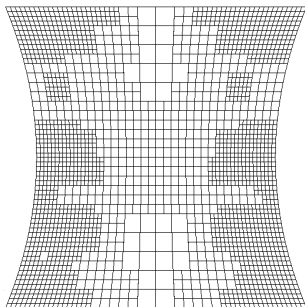
2d+t:

- $$u(x, t) = \sin(\pi x) \sin(\pi y) \sin(\pi t),$$
- $$f(x, t) = (\pi \sin(\pi x) \sin(\pi y)) (\cos(\pi t) + 2 \pi \sin(\pi t))$$



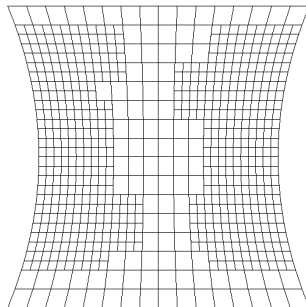
Example 2-1. Mesh refinement for $M_{BULK}(0.4)$ and $M_{BULK}(0.6)$

$M_{BULK}(0.4)$



ref. 5: \mathcal{K}_h

$M_{BULK}(0.6)$

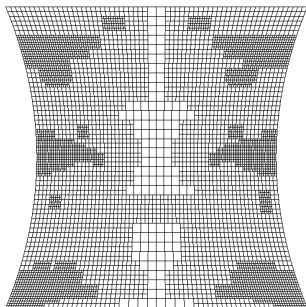


ref. 5: \mathcal{K}_h

$$u_h \in S_h^2, \mathbf{y}_h \in \oplus^2 S_{4h}^4, \text{ and } w_h \in S_{4h}^4.$$

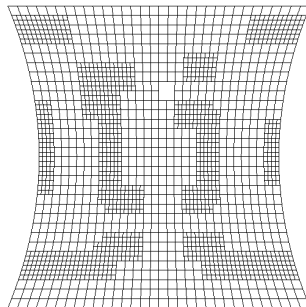
Example 2-1. Mesh refinement for $M_{BULK}(0.4)$ and $M_{BULK}(0.6)$

$M_{BULK}(0.4)$



ref. 6: \mathcal{K}_h

$M_{BULK}(0.6)$

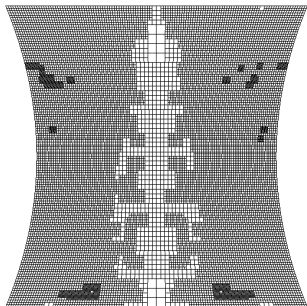


ref. 6: \mathcal{K}_h

$$u_h \in S_h^2, \mathbf{y}_h \in \oplus^2 S_{4h}^4, \text{ and } w_h \in S_{4h}^4.$$

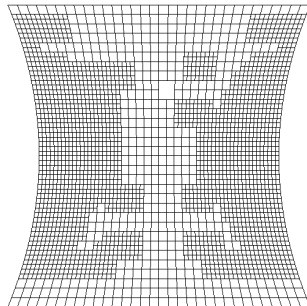
Example 2-1. Mesh refinement for $M_{BULK}(0.4)$ and $M_{BULK}(0.6)$

$M_{BULK}(0.4)$



ref. 7: \mathcal{K}_h

$M_{BULK}(0.6)$

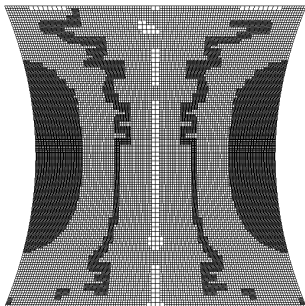


ref. 7: \mathcal{K}_h

$$u_h \in S_h^2, \mathbf{y}_h \in \oplus^2 S_{4h}^4, \text{ and } w_h \in S_{4h}^4.$$

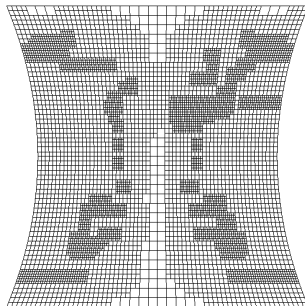
Example 2-1. Mesh refinement for $M_{BULK}(0.4)$ and $M_{BULK}(0.6)$

$M_{BULK}(0.4)$



ref. 8: \mathcal{K}_h

$M_{BULK}(0.6)$

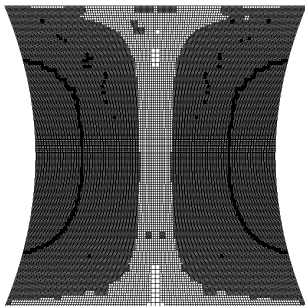


ref. 8: \mathcal{K}_h

$$u_h \in S_h^2, \mathbf{y}_h \in \oplus^2 S_{4h}^4, \text{ and } w_h \in S_{4h}^4.$$

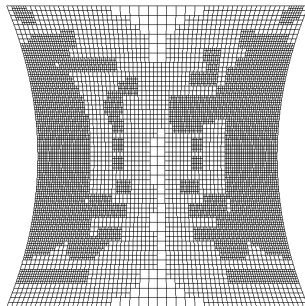
Example 2-1. Mesh refinement for $M_{BULK}(0.4)$ and $M_{BULK}(0.6)$

$M_{BULK}(0.4)$



ref. 9: \mathcal{K}_h

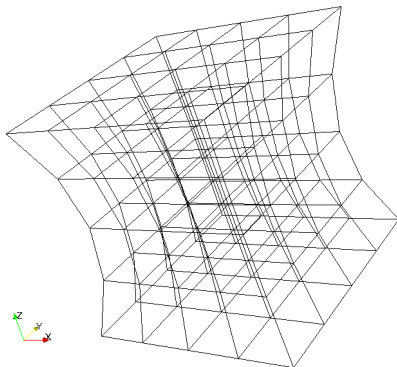
$M_{BULK}(0.6)$



ref. 9: \mathcal{K}_h

$$u_h \in S_h^2, \mathbf{y}_h \in \oplus^2 S_{4h}^4, \text{ and } w_h \in S_{4h}^4.$$

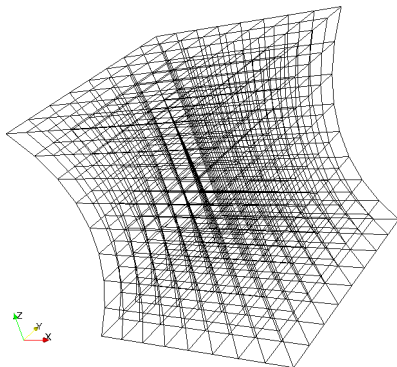
Example 2-2. Mesh refinement for $M_{BULK}(0.4)$



ref. 1: \mathcal{K}_h

$$u_h \in S_h^2, \mathbf{y}_h \in \oplus^2 S_{3h}^4, \text{ and } w_h \in S_{3h}^4.$$

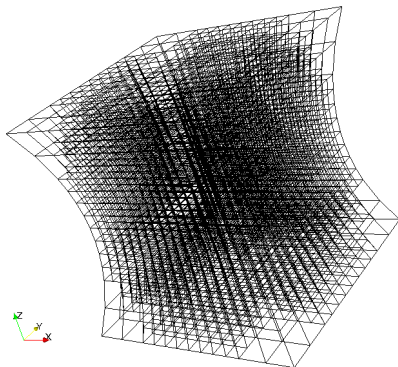
Example 2-2. Mesh refinement for $M_{BULK}(0.4)$



ref. 2: \mathcal{K}_h

$$u_h \in S_h^2, \mathbf{y}_h \in \oplus^2 S_{3h}^4, \text{ and } w_h \in S_{3h}^4.$$

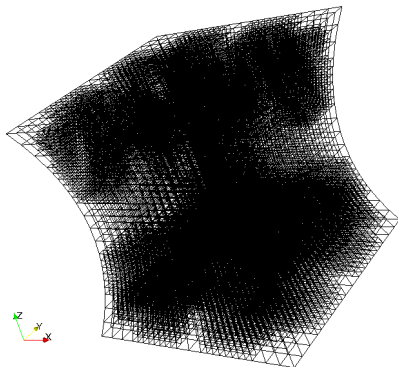
Example 2-2. Mesh refinement for $M_{BULK}(0.4)$



ref. 3: \mathcal{K}_h

$$u_h \in S_h^2, \mathbf{y}_h \in \oplus^2 S_{3h}^4, \text{ and } w_h \in S_{3h}^4.$$

Example 2-2. Mesh refinement for $M_{BULK}(0.4)$



ref. 4: \mathcal{K}_h

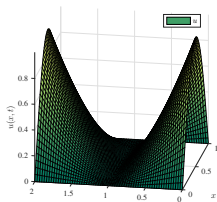
$$u_h \in S_h^2, \mathbf{y}_h \in \oplus^2 S_{3h}^4, \text{ and } w_h \in S_{3h}^4.$$

Example 3: Robustness to problems different singularities

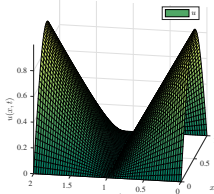
Given data:

- $\Omega = (0, 1), T = 2$

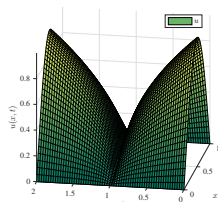
- $u(x, t) = \sin \pi x (1 - t)^\lambda, \lambda = \left\{ \frac{3}{2}, 1, \frac{1}{2} \right\}, u \in H^{s, \lambda + \frac{1}{2} - \varepsilon}(Q),$
where $\varepsilon > 0$ is arbitrary small, $s \geq 2$



(a) $u = \sin \pi x (1 - t)^{3/2}$



(b) $u = \sin \pi x |1 - t|$



(c) $u = \sin \pi x (1 - t)^{1/2}$

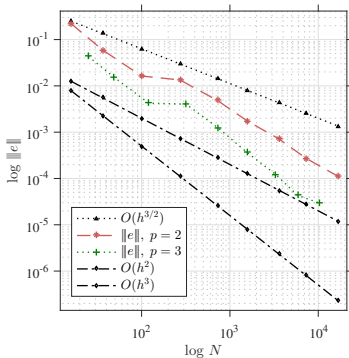
Discretisation:

- $u_h \in S_h^2$

- $y_h \in \oplus^2 S_h^3$ and $w_h \in S_h^3$

Example 3: Error order of convergence, $\lambda = \frac{3}{2}$

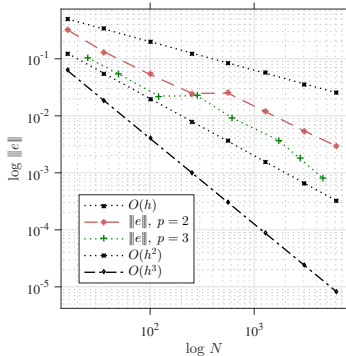
Theoretical (expected) rate $O(h^{3/2})$:



Error order of convergence for approximations with $u \in S_h^2$ and $u \in S_h^3$.

Example 3: Error order of convergence, $\lambda = 1$

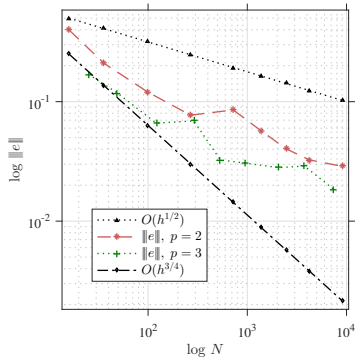
Theoretical (expected) rate $O(h)$:



Error order of convergence for approximations with $u \in S_h^2$ and $u \in S_h^3$.

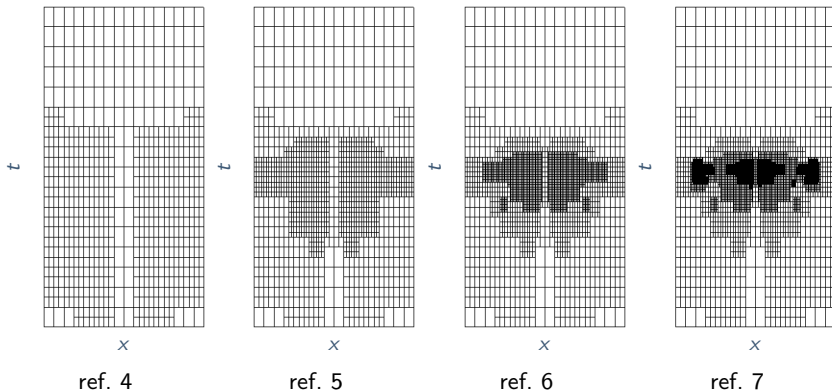
Example 3: Error order of convergence, $\lambda = \frac{1}{2}$

Theoretical (expected) rate $O(h^{1/2})$:



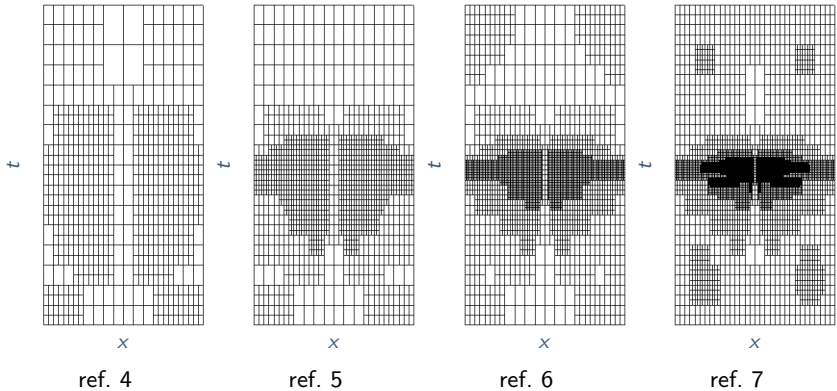
Error order of convergence for approximations with $u \in S_h^2$ and $u \in S_h^3$.

Example 3: Mesh refinement with $\mathbb{M}_{\text{BULK}}(0.4)$, $\lambda = \frac{3}{2}$



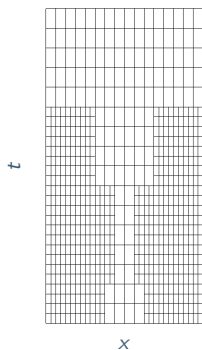
Meshes obtained on the refinement steps 4–7 for $u_h \in S_h^2$.

Example 3: Mesh refinement with $\mathbb{M}_{\text{BULK}}(0.4)$, $\lambda = 1$

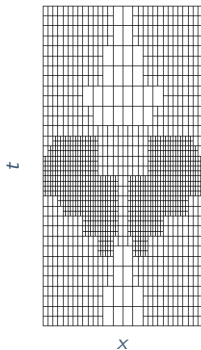


Meshes obtained on the refinement steps 4–7 for $u_h \in S_h^2$.

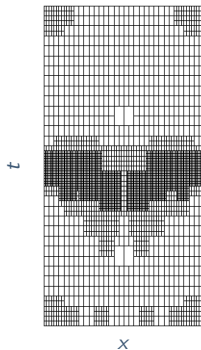
Example 3: Mesh refinement with $\mathbb{M}_{\text{BULK}}(0.4)$, $\lambda = \frac{1}{2}$



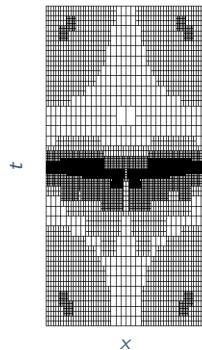
ref. 4



ref. 5



ref. 6



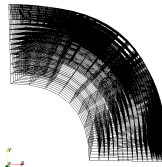
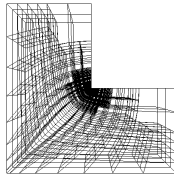
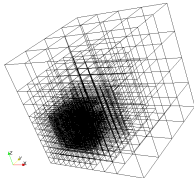
ref. 7

Meshes obtained on the refinement steps 4–7 for $u_h \in S_h^2$.

Conclusions and roadmap

Conclusions and roadmap

- From globally to locally stabilized space-time IgA schemes:
 - 1 a priori discretization error estimates
 - 2 functional a posteriori discretization error estimates
 - 3 adaptive IgA schemes based on global flux reconstruction
- Adaptivity + Fast (multilevel) solvers + Parallelization
 - 1 improving assembling time, in particular, for the THB-splines
 - 2 fast solvers for the system providing optimal flux used in the majorant



THANK YOU FOR YOUR ATTENTION!

U. Langer, S. Matculevich, and S. Repin. Adaptive Space-Time Isogeometric Analysis for Parabolic Evolution Problems, [arXiv.org](https://arxiv.org/abs/math.NA/1807.05950), math.NA:1807.05950, 2018 (submitted).

U. Langer, S. Matculevich, and S. Repin. Guaranteed error control bounds for the stabilized space-time IgA approximations to parabolic problems, [arXiv.org](https://arxiv.org/abs/math.NA/1712.06017), math.NA/1712.06017, 2017 (submitted).

U. Langer, S. Matculevich, and S. Repin. Functional type error control for Stabilized space- time IgA approximations to parabolic problems. In I. Lirkov and S. Margenov, editors, Large-Scale Scientific Computing (LSSC 2017), Lecture Notes in Computer Science (LNCS), p. 57–66, Springer-Verlag, 2017.

S. Matculevich. Functional approach to the error control in adaptive IgA schemes for elliptic boundary value problems *Journal of Computational and Applied Mathematics*, doi.org/10.1016/j.cam.2018.05.029, 2018.

U. Langer, S. Matculevich, and S. Repin. A posteriori error estimates for space-time IgA approximations to parabolic initial boundary value problems, [arXiv.org](https://arxiv.org/abs/math.NA/1612.08998), math.NA/1612.08998, 2016.

Example 1. Comparison of different approaches for flux reconstructions

# ref.	$\ e\ _Q$	$l_{\text{eff}}(\overline{M}^1)$	e.o.c. ($\ e\ _{s,h}$)
uniform refinement			expected $O(h^2)$
(a) majorant with $y_h := \operatorname{argmin}_{y_h \in Y_h} \overline{M}$			
2	2.5516e-03	1.07	3.44
4	1.5947e-04	1.39	2.36
6	9.9670e-06	1.00	2.09
8	6.2294e-07	1.01	2.02
(b) majorant with $y_h = \nabla_x u_h$ (implies residual-type estimate)			
2	2.5516e-03	4.52	3.44
4	1.5947e-04	1.91	2.36
6	9.9670e-06	4.09	2.09
8	6.2294e-07	6.39	2.02
(c) majorant with equilibrated fluxes (implies equalibration-type estimate)			
2	2.5516e-03	36.22	3.44
4	1.5947e-04	869.73	2.36
7	2.4918e-06	138992.85	2.05
8	6.2294e-07	247643.30	2.02

Efficiency of \overline{M}^1 w.r.t. three approaches of the $y_h \in \oplus^2 S_{5h}^3$ reconstruction.

Example 1. Comparison of different approaches for flux reconstructions

# ref.	$\ e\ _Q$	$l_{\text{eff}}(\bar{M}^1)$	e.o.c. ($\ e\ _{S,h}$)
adaptive refinement ($\sigma = 0.4$)			
majorant with $y_h := \operatorname{argmin}_{y_h \in Y_h} \bar{M}$			
2	2.5516e-03	1.07	3.42
4	2.2734e-04	1.41	2.36
6	2.9493e-05	1.08	2.70
8	4.8121e-06	1.12	1.56
majorant with $y_h = \nabla_x u_h$ (implies residual-type estimate)			
2	2.5516e-03	4.52	3.42
4	2.2734e-04	1.80	1.49
6	2.6218e-05	3.57	2.43
8	3.1014e-06	3.43	2.74
majorant with equilibrated fluxes (implies equilibration-type estimate)			
2	2.5516e-03	30.14	3.42
4	2.1893e-04	705.39	2.10
6	3.7533e-05	3663.95	2.78
8	1.0382e-05	33422.45	2.11

Efficiency of \bar{M}^1 w.r.t. three approaches of the $y_h \in \oplus^2 S_{5h}^3$ reconstruction.

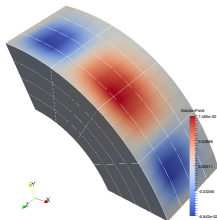
Example 4. Robustness to non-trivial domains

Given data:

- Ω is a quarter-annulus,
 $T = 1$
- $u = (1-x)x^2(1-y)y^2(1-t)t^2$
- $f = \dots$
- $u_D = u$

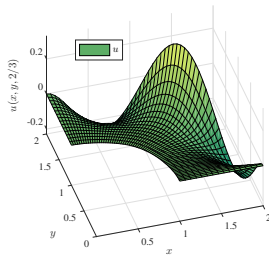
Discretisation:

- $u_h \in S_h^2$
- $y_h \in \oplus^2 S_{2h}^3$
- $w_h \in S_{2h}^3$



(a)

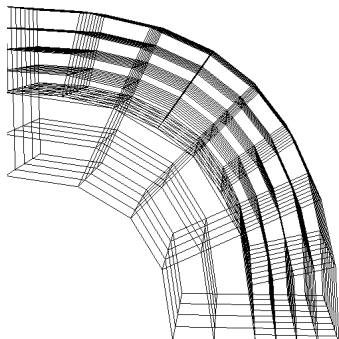
(a) u in 2d+t axis.



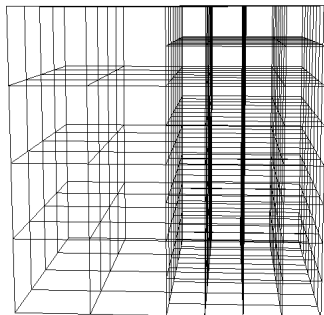
(b)

(b) u at $t = \frac{2}{3}$.

Example 4. Meshes on parametric and physical domains for $M_{BULK}(0.6)$

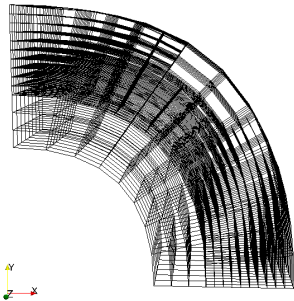


ref. 1: \hat{Q} and $\hat{\mathcal{K}}_h$

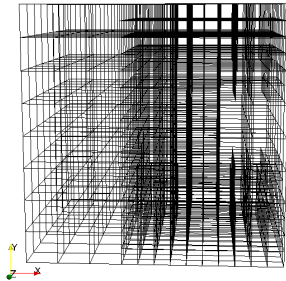


ref. 1: Q and \mathcal{K}_h

Example 4. Meshes on parametric and physical domains for $M_{BULK}(0.6)$

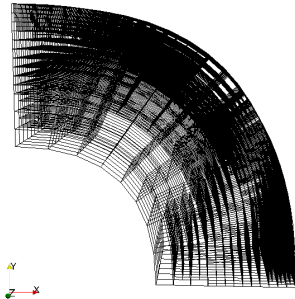


ref. 2: \hat{Q} and $\hat{\mathcal{K}}_h$

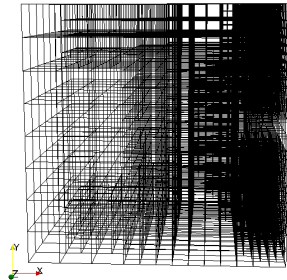


ref. 2: Q and \mathcal{K}_h

Example 4. Meshes on parametric and physical domains for $M_{BULK}(0.6)$



ref. 3: \hat{Q} and \hat{K}_h

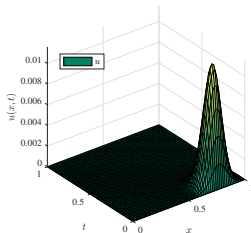


ref. 3: Q and K_h

Example 3. Robustness to solutions with sharp local Gaussian jumps

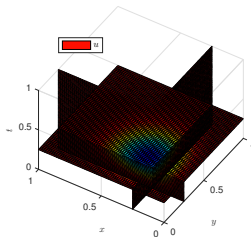
Example 3-1: 1d + t

- $\Omega = (0, 1), T = 1$
- $u = (x^2 - x)(t^2 - t) e^{-100|(x,t)-(0.8,0.05)|}$
- $f = \dots$
- $u_D = 0$



Example 3-2: 2d + t

- $\Omega = (0, 1)^2, T = 2$
- $u = (x^2 - x)(y^2 - y)(t^2 - t) e^{-100|(x,y,t)-(0.25,0.25,0.25)|}$
- $f = \dots$
- $u_D = 0$

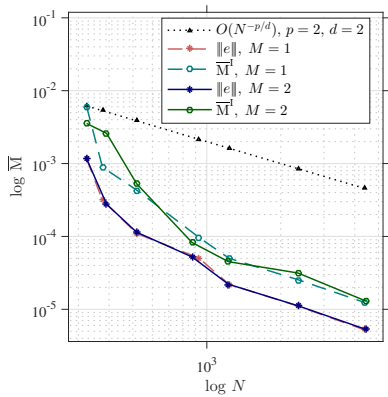


Example 3. Adaptive refinement, $u_h \in S_h^2$

# ref.	$\ e\ _Q$	$I_{\text{eff}}(\bar{M}^I)$	$I_{\text{eff}}(\bar{M}^{II})$	$\ e\ _{loc,h}$	$\ e\ _{\mathcal{L}}$	$I_{\text{eff}}(\mathbb{E}I)$	e.o.c. ($\ e\ _{loc,h}$)	e.o.c. ($\ e\ _{\mathcal{L}}$)
(a) $u_h \in S_h^2, y_h \in \oplus^2 S_h^3, \text{ and } w_h \in S_h^3$								
2	3.1311e-04	2.85	1.55	3.1335e-04	5.6510e-02	1.00	17.71	8.64
3	1.0915e-04	3.93	1.73	1.0944e-04	3.1506e-02	1.00	6.49	3.60
5	2.2033e-05	2.27	1.36	2.2042e-05	1.4796e-02	1.00	5.87	3.59
7	5.2517e-06	2.38	1.22	5.2526e-06	7.2473e-03	1.00	2.41	1.27
(b) $u_h \in S_h^2, y_h \in \oplus^2 S_{2h}^6, \text{ and } w_h \in S_{2h}^6$								
2	2.7623e-04	9.39	2.38	2.7647e-04	5.4452e-02	1.00	15.35	7.24
3	1.1419e-04	4.62	1.79	1.1446e-04	3.1695e-02	1.00	6.27	3.85
5	2.2089e-05	2.04	1.13	2.2099e-05	1.4911e-02	1.00	5.07	3.04
7	5.2825e-06	2.45	1.24	5.2837e-06	7.1577e-03	1.00	2.35	1.24

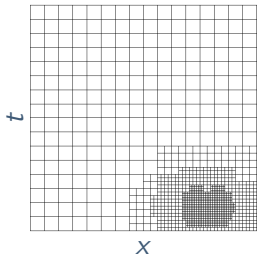
Efficiency of $\bar{M}^I, \bar{M}^{II}, \bar{M}_{s,h}^I$, and $\mathbb{E}I$.

Example 3. Error order of convergence for $u_h \in S_h^2$

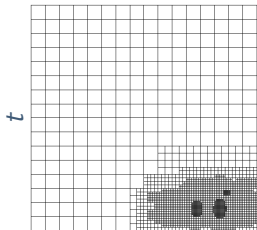


The majorant and e.o.c. for (a) $y_h \in \oplus^2 S_h^3$, and $w_h \in S_h^3$ and (b) $y_h \in \oplus^2 S_{2h}^6$, and $w_h \in S_{2h}^6$.

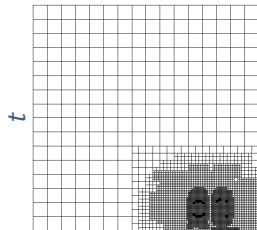
Example 3-1. Mesh refinement for $M_{BULK}(0.6)$



ref. 4



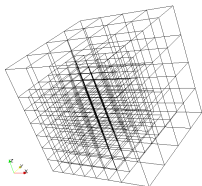
ref. 5



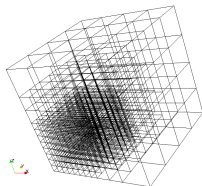
ref. 6

Mesheres obtained on the refinement steps 4–6.

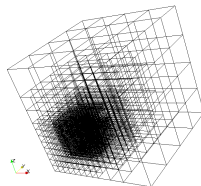
Example 3-2. Mesh refinement for $M_{BULK}(0.6)$



ref. 1: Q and \mathcal{K}_h



ref. 2: Q and \mathcal{K}_h



ref. 3: Q and \mathcal{K}_h

Meshes obtained on the refinement steps 1–3 $u_h \in S_h^2$, $\mathbf{y}_h \in \oplus^2 S_h^3$ and $w_h \in S_h^3$