

Numerical Algorithms for Algebraic Stabilizations of Scalar Convection-Dominated Problems

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Applied and Computational Mathematics

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1 Introduction

2 Iteration Schemes

3 A Posteriori Error Analysis

3.1 Residual Based Approach

3.2 AFC-SUPG Approach

4 Hanging nodes

5 Conclusions

- Steady-state convection-diffusion-reaction equation

$$\begin{aligned} -\varepsilon \Delta u + \mathbf{b} \cdot \nabla u + cu &= f && \text{in } \Omega \\ u &= u^b && \text{on } \Gamma, \end{aligned} \tag{1}$$

- Ω – bounded polyhedral Lipschitz domain in \mathbb{R}^d , $d \in \{2, 3\}$
- Assume

$$\left(c(\mathbf{x}) - \frac{1}{2} \nabla \cdot \mathbf{b}(\mathbf{x}) \right) \geq \sigma > 0$$

- Interested in **convection-dominated regime**, $\varepsilon \ll \|\mathbf{b}\|_{L^\infty(\Omega)}$

- Ideal discretization

1. Sharp layers

- Many discretizations satisfy this property, e.g., SUPG¹
- Reasonably well for AFC schemes

¹Brooks, Hughes: CMAME (32), 199-259, 1982

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 - Usually not satisfied for nonlinear discretizations, like AFC schemes
- Because of 2nd property: AFC schemes very well suited for applications

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- Hemker Problem

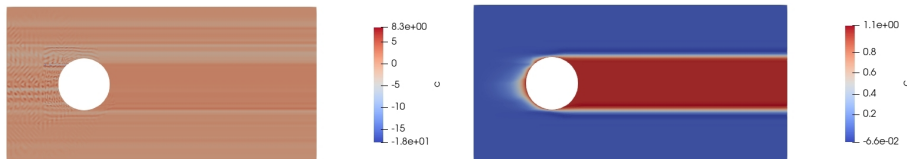


Figure 1: Galerkin Solution and SUPG Solution



Figure 2: AFC Solution

- Derivation
 - Galerkin FEM (Algebraic form)

$$\sum_{j=1}^N a_{ij} u_j = f_i, \quad i = 1, \dots, M,$$
$$u_i = u_i^b, \quad i = M + 1, \dots, N$$

- **Derivation**

- Galerkin FEM (Algebraic form)

$$\sum_{j=1}^N a_{ij} u_j = f_i, \quad i = 1, \dots, M,$$
$$u_i = u_i^b, \quad i = M + 1, \dots, N$$

- Artificial diffusion matrix D

$$d_{ij} = d_{ji} = -\max\{a_{ij}, 0, a_{ji}\} \quad \forall i \neq j, \quad d_{ii} = -\sum_{i \neq j} d_{ij}$$

- Anti-diffusive fluxes

$$f_{ij} = d_{ij}(u_j - u_i), \quad f_{ij} = -f_{ji}, \quad i, j = 1, \dots, N$$

- Derivation (cont.)
 - Solution-dependent coefficients

$$\alpha_{ij} = \alpha_{ji}, \quad i, j = 1, \dots, N$$

with

$$\alpha_{ij} \in [0, 1]$$

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- Final scheme

$$\sum_{j=1}^N a_{ij} u_j + \sum_{j=1}^N (1 - \alpha_{ij}) d_{ij} (u_j - u_i) = f_i, \quad i = 1, \dots, M,$$

$$u_i = u_i^b, \quad i = M + 1, \dots, N$$

- Limiters

- Kuzmin limiter^{1,2,3}
- BJK limiter⁴
- Analytical properties^{2,3}
- BJK limiter in general more accurate⁴

¹ Kuzmin: in Proc. Int. Conf. Comput. Meth. for Coupled Problems in Science and Engineering, CIMNE, 2007

² Barrenechea, John, Knobloch: SINUM 54, 2427–2451, 2016

³ Barrenechea, John, Knobloch, Rankin: SeMA Journal 75, 655–685, 2018

⁴ Barrenechea, John, Knobloch: M3AS 27, 525–548, 2017

- Fixed point iteration with changing matrix

$$\sum_{j=1}^N a_{ij} u_j^{(m+1)} + \sum_{j=1}^N \left(1 - \alpha_{ij}^{(m)}\right) d_{ij} \left(u_j^{(m+1)} - u_i^{(m+1)}\right) = f_i,$$
$$u_i^{(m+1)} = u_i^b$$

- Fixed point iteration with fixed matrix: using

$$\sum_{j=1}^N (1 - \alpha_{ij}) d_{ij} (u_j - u_i) = \sum_{j=1}^N d_{ij} u_j - u_i \underbrace{\sum_{j=1}^N d_{ij}}_{=0} - \sum_{j=1}^N \alpha_{ij} d_{ij} (u_j - u_i),$$

gives

$$\sum_{j=1}^N (a_{ij} + d_{ij}) u_j^{(m+1)} = f_i + \sum_{j=1}^N \alpha_{ij}^{(m)} f_{ij}^{(m)}, \quad i = 1, \dots, M,$$
$$u_i^{(m+1)} = u_i^b, \quad i = M + 1, \dots, N$$

- Fixed point iterations
 - Fixed point iteration with fixed matrix (FPR)
 - Matrix is M-matrix
 - With direct sparse solver: factorization is needed only once
 - Fixed point iteration with changing matrix (FPM)
 - More implicit approach, hope for better convergence properties
 - General fixed point iteration by linear combination

$$\begin{aligned} & \sum_{j=1}^N (a_{ij} + d_{ij}) u_j^{(m+1)} - \omega_{\text{fp}} \sum_{j=1}^N \alpha_{ij}^{(m)} d_{ij} (u_j^{(m+1)} - u_i^{(m+1)}) \\ &= f_i + (1 - \omega_{\text{fp}}) \sum_{j=1}^N \alpha_{ij}^{(m)} f_{ij}^{(m)}, \quad i = 1, \dots, M, \\ u_i^{(m+1)} &= u_i^b, \quad i = M + 1, \dots, N \end{aligned}$$

- Formal Newton method
 - Formal derivation of Jacobian

$$DF\left(u^{(m)}\right)_{ij} = \begin{cases} a_{ij} + d_{ij} - \alpha_{ij}^{(m)} d_{ij} - \sum_{k=1}^N \frac{\partial \alpha_{ik}^{(m)}}{\partial u_j} d_{ik} \left(u_k^{(m)} - u_i^{(m)}\right) & \text{if } i \neq j, \\ a_{ii} + d_{ii} + \sum_{\substack{j=1 \\ j \neq i}}^N \alpha_{ij}^{(m)} d_{ij} - \sum_{k=1}^N \frac{\partial \alpha_{ik}^{(m)}}{\partial u_i} d_{ik} \left(u_k^{(m)} - u_i^{(m)}\right) & \text{if } i = j \end{cases}$$

- Formal Newton method: how to deal with non-smooth cases?
- Discussion only for Kuzmin limiter
 - Involves maxima and minima of two arguments, one of them is constant
 - 1 Non-regularized approach
 - Take one-sided derivative w.r.t. constant, i.e., set value to zero
 - 2 Regularized approach
 - Replace maximum for some $\sigma_N > 0$ by¹

$$\max_{\sigma_N(x,y)} = \frac{1}{2} \left(x + y + \sqrt{(x-y)^2 + \sigma_N} \right)$$

- We did not regularize the limiter in the equation, only in the iteration matrix

¹Badia, Bonilla: CMAME 313, 133–158, 2017

- General form of the matrix

$$\underbrace{\underbrace{a_{ij} + d_{ij}}_{\text{FPR, const. matrix}} - \omega_{\text{fp}} \alpha_{ij} d_{ij} + \omega_{\text{jac}} (\text{term with der. of } \alpha_{ij})}_{\text{FPM, changing matrix}}, \quad i \neq j$$

formal Newton

- Similar for diagonal entries
- Some modifications for regularized Newton approach
- Iteration

$$\mathbf{u}^{(m+1)} = \mathbf{u}^{(m)} + \omega \left(\mathbf{u}^{(m+1)} - \mathbf{u}^{(m)} \right)$$

- Algorithmic components
 - Adaptive choice of damping parameter ¹
 - Anderson acceleration ²
 - Projection to admissible values ³
 - Selection of initial iterate ⁴

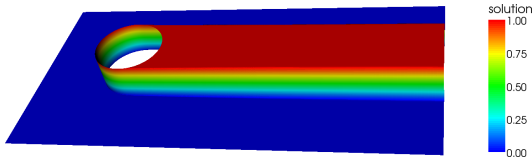
¹ John, Knobloch: CMAME 197, 1997–2014, 2008

² Walker, Ni: SINUM 49(4), 1715–1735, 2011

³ Badia, Bonilla: CMAME 313, 133–158, 2017

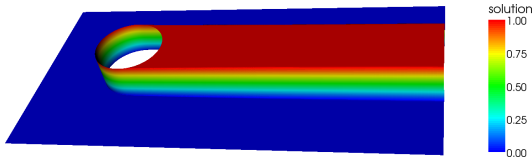
⁴ J., John: BAIL 2018- Boundary and Interior Layers, (in press)

- Hemker problem¹
- Various values of $\varepsilon = 10^{-6}$, $\mathbf{b} = (1, 0)^T$, $\mathbf{c} = \mathbf{f} = 0$



¹Hemker: JCAM 76, 277-285, 1996

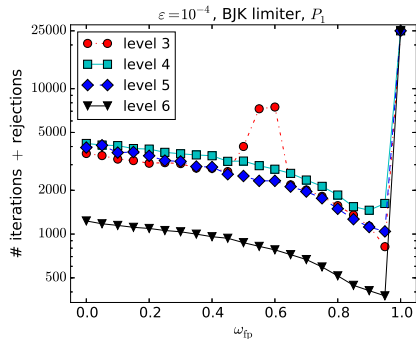
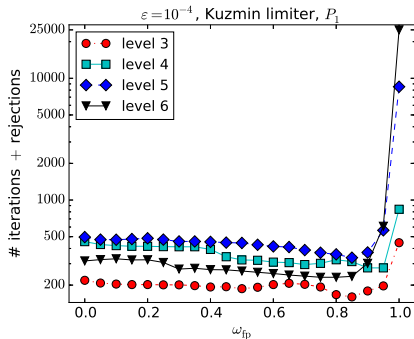
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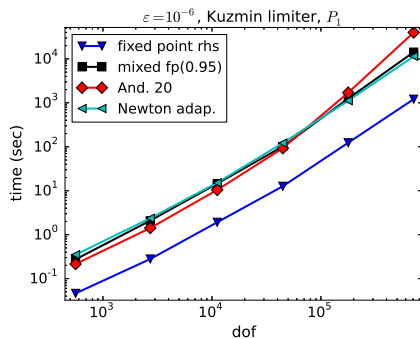
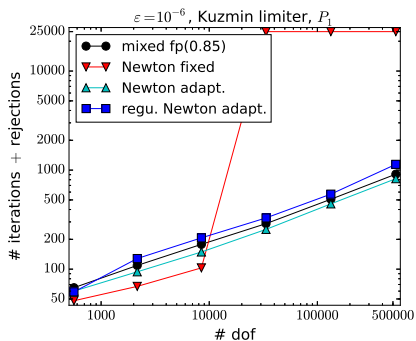
- P_1 finite elements
- Stopping criteria
 - $\|\text{residual}\|_2 \leq \sqrt{\#\text{dof}} 10^{-10}$
 - 25000 iterations
- Direct sparse solver (UMFPACK)

¹Hemker: JCAM 76, 277-285, 1996

● Hemker problem¹



¹Hemker: JCAM 76, 277-285, 1996



- FPR most efficient in 3d

- **Variational problem** of AFC scheme reads
Find $u_h \in V_h$ such that

$$a_h(u_h, v_h) + d_h(u_h; u_h, v_h) = \langle f, v_h \rangle \quad \forall v_h \in V_h$$

- V_h – finite element space with homogeneous Dirichlet boundary conditions ($V_h \subset V$)
- stabilization

$$d_h(w_h; z_h, v_h) = \sum_{i,j=1}^N (1 - \alpha_{ij}(w_h)) d_{ij}(z_j - z_i) v_i \quad \forall w_h, v_h, z_h \in V_h$$

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- Another representation of stabilization for $w_h, v_h, z_h \in V_h$,¹

$$d_h(w_h; z_h, v_h) = \sum_{E \in \mathcal{E}_h} (1 - \alpha_E(w_h)) d_E h_E (\nabla z_h \cdot \mathbf{t}_E, \nabla v_h \cdot \mathbf{t}_E)$$

¹Barrenechea, John, Knobloch, Rankin: SeMA Journal 75, 655–685, 2018

- AFC norm

$$\|u_h\|_{\text{AFC}}^2 = \|u_h\|_a^2 + d_h(u_h, u_h, u_h) \quad \forall u_h \in V_h$$

- where $\|u_h\|_a^2 = \varepsilon |u_h|_1^2 + \sigma \|u_h\|_0^2$

- Let $I_h u$ denote an interpolation operator. Galerkin orthogonality arguments

$$\begin{aligned} \|u - u_h\|_{\text{AFC}}^2 &= \langle f, u - I_h u \rangle + \langle g, u - I_h u \rangle_{\Gamma_N} - a(u_h, u - I_h u) \\ &\quad + d_h(u_h; u, I_h u - u_h) \end{aligned}$$

¹John, Novo: CMAME 255, 289–305, 2013

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- Standard residual posteriori error bound ¹

$$\begin{aligned} &\langle f, u - I_h u \rangle + \langle g, u - I_h u \rangle_{\Gamma_N} - a(u_h, u - I_h u) \\ &= \sum_{K \in \mathcal{T}_h} (R_K(u_h), u - I_h u)_K + \sum_{F \in \mathcal{F}_h} \langle R_F(u_h), u - I_h u \rangle_F \end{aligned}$$

¹ John, Novo: CMAME 255, 289–305, 2013

with

$$\begin{aligned} R_K(\mathbf{u}_h) &:= f + \varepsilon \Delta \mathbf{u}_h - \mathbf{b} \cdot \nabla \mathbf{u}_h - c \mathbf{u}_h|_K, \\ R_F(\mathbf{u}_h) &:= \begin{cases} -\varepsilon [|\nabla \mathbf{u}_h \cdot \mathbf{n}_F|]_F & \text{if } F \in \mathcal{F}_{h,\Omega}, \\ \mathbf{g} - \varepsilon (\nabla \mathbf{u}_h \cdot \mathbf{n}_F) & \text{if } F \in \mathcal{F}_{h,N}, \\ 0 & \text{if } F \in \mathcal{F}_{h,D} \end{cases} \end{aligned}$$

with

$$R_K(u_h) := f + \varepsilon \Delta u_h - \mathbf{b} \cdot \nabla u_h - cu_h|_K,$$
$$R_F(u_h) := \begin{cases} -\varepsilon [|\nabla u_h \cdot \mathbf{n}_F|]_F & \text{if } F \in \mathcal{F}_{h,\Omega}, \\ g - \varepsilon (\nabla u_h \cdot \mathbf{n}_F) & \text{if } F \in \mathcal{F}_{h,N}, \\ 0 & \text{if } F \in \mathcal{F}_{h,D} \end{cases}$$

- Using interpolation estimates, Cauchy-Schwarz and Young's inequality

$$\begin{aligned} & \|u - u_h\|_a^2 + \frac{C_Y}{C_Y - 1} d_h(u_h; u - u_h, u - u_h) \\ & \leq \frac{C_Y^2}{2(C_Y - 1)} \sum_{K \in \mathcal{T}_h} \min \left\{ \frac{C_I^2}{\sigma}, \frac{C_I^2 h_K^2}{\varepsilon} \right\} \|R_K(u_h)\|_{L^2(K)}^2 \\ & \quad + \frac{C_Y^2}{2(C_Y - 1)} \sum_{F \in \mathcal{F}_h} \min \left\{ \frac{C_F^2 h_F}{\varepsilon}, \frac{C_F^2}{\sigma^{1/2} \varepsilon^{1/2}} \right\} \|R_F(u_h)\|_{L^2(F)}^2 \\ & \quad + \frac{C_Y}{C_Y - 1} d_h(u_h; u, I_h u - u_h) \end{aligned}$$

- Linearity of $d_h(\cdot; \cdot, \cdot)$,

$$d_h(u_h; u, I_h u - u_h) = d_h(u_h; u - u_h, I_h u - u_h) + d_h(u_h; u_h, I_h u - u_h)$$

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- Using interpolation estimates, Cauchy-Schwarz, trace inequality, inverse estimate, and Young's inequality

$$\begin{aligned} d_h(u_h; u_h, I_h u - u_h) &\leq \frac{C_Y}{2} \sum_{E \in \mathcal{E}_h} \min \left\{ \frac{\kappa_1 h_E^2}{\varepsilon}, \frac{\kappa_2}{\sigma} \right\} (1 - \alpha_E)^2 |d_E|^2 h_E^{1-d} \\ &\quad \times \|\nabla u_h \cdot \mathbf{t}_E\|_{L^2(E)}^2 + \frac{1}{C_Y} \|u - u_h\|_a^2, \end{aligned}$$

where

$$\begin{aligned} \kappa_1 &= C_{\text{edge,max}}(1 + (1 + C_I)^2), \\ \kappa_2 &= C_{\text{inv}}^2 C_{\text{edge,max}}(1 + (1 + C_I)^2) \end{aligned}$$

Theorem [Global a posteriori error estimate] A global a posteriori error estimate for the energy norm is given by

$$\|u - u_h\|_a^2 \leq \eta_1^2 + \eta_2^2 + \eta_3^2,$$

where

$$\eta_1^2 = \sum_{K \in \mathcal{T}_h} \min \left\{ \frac{4C_I^2}{\sigma}, \frac{4C_I^2 h_K^2}{\varepsilon} \right\} \|R_K(u_h)\|_{L^2(K)}^2,$$

$$\eta_2^2 = \sum_{F \in \mathcal{F}_h} \min \left\{ \frac{4C_F^2 h_F}{\varepsilon}, \frac{4C_F^2}{\sigma^{1/2} \varepsilon^{1/2}} \right\} \|R_F(u_h)\|_{L^2(F)}^2,$$

$$\eta_3^2 = \sum_{E \in \mathcal{E}_h} \min \left\{ \frac{4\kappa_1 h_E^2}{\varepsilon}, \frac{4\kappa_2}{\sigma} \right\} (1 - \alpha_E)^2 |d_E|^2 h_E^{1-d} \|\nabla u_h \cdot \mathbf{t}_E\|_{L^2(E)}^2.$$

- Local lower bound for a mesh cell K

$$\eta_K^2 = \eta_{\text{int},K}^2 + \sum_{F \in \mathcal{F}_h(K)} \eta_{\text{Face},F}^2 + \sum_{E \in \mathcal{E}_h(K)} \eta_{d_h,E}^2$$

- Local lower bound for a mesh cell K

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where

$$\eta_{\text{Int},K}^2 = \min \left\{ \frac{4C_I^2}{\sigma}, \frac{4C_I^2 h_K^2}{\varepsilon} \right\} \|R_{K,h}(u_h)\|_{L^2(K)}^2,$$

$$\eta_{\text{Face},F}^2 = \frac{1}{2} \min \left\{ \frac{4C_F^2 h_F}{\varepsilon}, \frac{4C_F^2}{\sigma_0^{1/2} \varepsilon^{1/2}} \right\} \|R_F(u_h)\|_{L^2(F)}^2,$$

$$\eta_{d_h,E}^2 = \min \left\{ \frac{4\kappa_1 h_E^2}{\varepsilon}, \frac{4\kappa_2}{\sigma} \right\} (1 - \alpha_E)^2 |d_E|^2 h_E^{1-d} \|\nabla u_h \cdot \mathbf{t}_E\|_{L^2(E)}^2$$

- Using standard bubble function arguments

$$\begin{aligned} \eta_{\text{Int},K} &\leq C \left(\max \left\{ C_K^2 + \frac{C_K h_K}{\varepsilon} \|\mathbf{b}\|_{L^\infty(K)}, \frac{C_K}{\sigma} \|c\|_{L^\infty(K)} \right\} \|u - u_h\|_{a,K} \right. \\ &\quad \left. + \frac{h_K}{\varepsilon^{1/2}} C_K \left(\|f - f_h\|_{0,K} + \|(\mathbf{b} - \mathbf{b}_h) \cdot \nabla u_h\|_{0,K} + \|(c - c_h)u_h\|_{0,K} \right) \right) \end{aligned}$$

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and

$$\eta_{\text{Face},F} \leq C \left(\max \left\{ C_F + \frac{C_F h_F \|\mathbf{b}\|_{L^\infty(\omega_F)}}{\varepsilon}, \frac{C_F h_F \|c\|_{L^\infty(\omega_F)}}{\varepsilon^{1/2} \sigma^{1/2}} \right\} \|u - u_h\|_{a,\omega_F} + \delta_{F \in \mathcal{F}_{h,N}} \frac{h_F^{1/2}}{\varepsilon^{1/2}} \|g - g_h\|_{L^2(F)} + \sum_{K \in \omega_F} \left[\eta_{\text{Int},K} + \frac{h_K}{\varepsilon^{1/2}} \left(\|f - f_h\|_{0,K} + \|(\mathbf{b} - \mathbf{b}_h) \cdot \nabla u_h\|_{0,K} + \|(c - c_h)u_h\|_{0,K} \right) \right] \right)$$

- For the stabilization term, from ¹ we get

$$|d_E| \leq C (\varepsilon + \|\mathbf{b}\|_{L^\infty(\Omega)} h + \|c\|_{L^\infty(\Omega)} h^2) h_E^{d-2}$$

¹Barrenechea, John, Knobloch, Rankin: SeMA Journal 75, 655–685, 2018

- For the stabilization term, from ¹ we get

$$|d_E| \leq C (\varepsilon + \|\mathbf{b}\|_{L^\infty(\Omega)} h + \|\mathbf{c}\|_{L^\infty(\Omega)} h^2) h_E^{d-2}$$

Hence,

$$\begin{aligned} \eta_{d_h, E} &\leq C \sum_{E \in \mathcal{E}_h} (1 - \alpha_E) (\varepsilon + \|\mathbf{b}\|_{L^\infty(\Omega)} h + \|\mathbf{c}\|_{L^\infty(\Omega)} h^2) \\ &\quad \times \frac{h_E^{(3-d)/2}}{\varepsilon^{1/2}} \|\nabla u_h \cdot \mathbf{t}_E\|_{L^2(E)} \end{aligned}$$

¹Barrenechea, John, Knobloch, Rankin: SeMA Journal 75, 655–685, 2018

Theorem [Local lower bound] There exists a constant $C > 0$, independent of the size of elements of \mathcal{T} , such that, for every $K \in \mathcal{T}$, the following local lower bound holds

$$\begin{aligned}
 & \eta_{\text{Int},K} + \sum_{K \in \mathcal{F}_h(K)} \eta_{\text{Face},F} + \sum_{E \in \mathcal{E}_h(K)} \eta_{d_h,E} \\
 & \leq \max \left\{ C_K^2 + \frac{C_K h_K}{\varepsilon} \|\mathbf{b}\|_{L^\infty(K)}, \frac{C_K}{\sigma_0} \|c\|_{L^\infty(K)} \right\} \|u - u_h\|_{a,\omega_K} \\
 & \quad + C \sum_{K \in \omega_K} \frac{h_K}{\varepsilon^{1/2}} \left(\|f - f_h\|_{0,K} + \|(\mathbf{b} - \mathbf{b}_h) \cdot \nabla u_h\|_{0,K} + \|(c - c_h)u_h\|_{0,K} \right) \\
 & \quad + C \sum_{F \in \mathcal{F}_h(K)} \delta_{F \in \mathcal{F}_{h,N}} \frac{h_F^{1/2}}{\varepsilon^{1/2}} \|g - g_h\|_{L^2(F)} \\
 & \quad + \sum_{E \in \mathcal{E}_h(K)} h^{1-d/2} \frac{h^{1/2}}{\varepsilon^{1/2}} \left(\varepsilon + \|\mathbf{b}\|_{L^\infty(\Omega)} h + \|c\|_{L^\infty(\Omega)} h^2 \right) \|\nabla u_h \cdot \mathbf{t}_E\|_{L^2(E)}.
 \end{aligned}$$

- The initial solution for the nonlinear loop is the SUPG solution ¹
 - u_{AFC} := AFC solution
 - u_{SUPG} := SUPG solution
- By triangle inequality

$$\begin{aligned}\|u - u_{\text{AFC}}\|_a^2 &\leq 2 \left(\|u - u_{\text{SUPG}}\|_a^2 + \|u_{\text{SUPG}} - u_{\text{AFC}}\|_a^2 \right) \\ &\leq 2 \left(\|u - u_{\text{SUPG}}\|_{\text{SUPG}}^2 + \|u_{\text{SUPG}} - u_{\text{AFC}}\|_a^2 \right)\end{aligned}$$

¹ J. John, BAIL 2018 (135), 2020

² John, Novo: CMAME, 255, 289-305, 2013

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- Using estimators from ²

$$\|u - u_{\text{SUPG}}\|_{\text{SUPG}}^2 \leq \eta_{\text{SUPG}}^2$$

¹ J. John, BAIL 2018 (135), 2020

² John, Novo: CMAME, 255, 289-305, 2013

- The initial solution for the nonlinear loop is the SUPG solution ¹
 - u_{AFC} := AFC solution
 - u_{SUPG} := SUPG solution
- By triangle inequality

$$\begin{aligned}\|u - u_{\text{AFC}}\|_a^2 &\leq 2 \left(\|u - u_{\text{SUPG}}\|_a^2 + \|u_{\text{SUPG}} - u_{\text{AFC}}\|_a^2 \right) \\ &\leq 2 \left(\|u - u_{\text{SUPG}}\|_{\text{SUPG}}^2 + \|u_{\text{SUPG}} - u_{\text{AFC}}\|_a^2 \right)\end{aligned}$$

- Using estimators from ²

$$\|u - u_{\text{SUPG}}\|_{\text{SUPG}}^2 \leq \eta_{\text{SUPG}}^2$$

and denoting

$$\eta_{\text{AFC-SUPG}}^2 := \|u_{\text{SUPG}} - u_{\text{AFC}}\|_a^2$$

⇒

$$\|u - u_h\|_a^2 \leq 2 \left(\eta_{\text{SUPG}}^2 + \eta_{\text{AFC-SUPG}}^2 \right)$$

¹J. John, BAIL 2018 (135), 2020

²John, Novo: CMAME, 255, 289-305, 2013

- Standard strategy for solving

SOLVE → **ESTIMATE** → **MARK** → **REFINE**

- Effectivity index for the estimator

$$\eta_{\text{eff}} = \frac{\eta}{\|u - u_h\|_a}$$

- Standard strategy for solving

SOLVE → **ESTIMATE** → **MARK** → **REFINE**

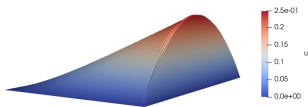
- Effectivity index for the estimator

$$\eta_{\text{eff}} = \frac{\eta}{\|u - u_h\|_a}$$

- Comparison of results:
 - On effectivity index
 - Adaptive grid refinement
 - Behavior of η_{d_h}
 - Behavior of η_{SUPG} and $\eta_{\text{AFC-SUPG}}$

- $\varepsilon = 10^{-3}$, $\mathbf{b} = (2, 1)^T$, $c = 1$ and f such that

$$u(x, y) = y(1 - y) \left(x - \frac{e^{(x-1)/\varepsilon} - e^{-1/\varepsilon}}{1 - e^{-1/\varepsilon}} \right)$$

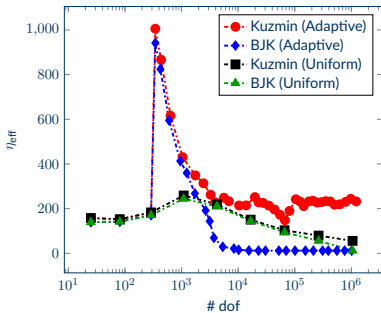


- proposed in ¹
- P_1 finite elements
- stop of the adaptive algorithm
 - $\eta \leq 10^{-3}$
 - #nDOFs $\approx 10^6$

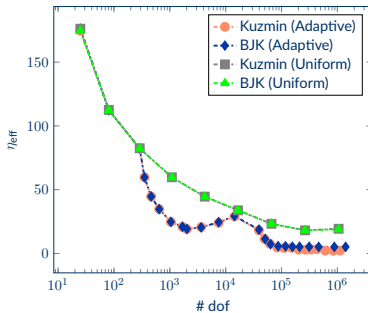
¹Allendes et. al, SISC, 39(5):A1903-A1927, 2017

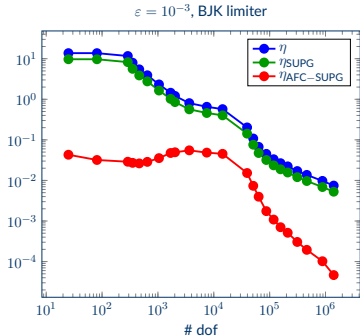
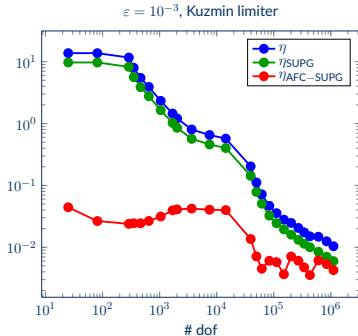
• Effectivity index

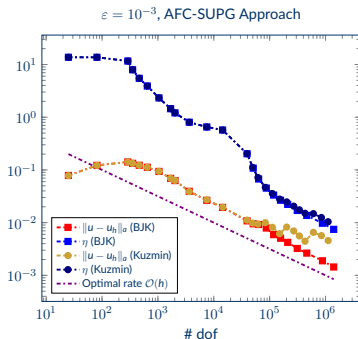
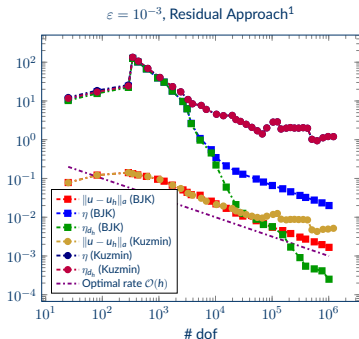
$\varepsilon = 10^{-3}$, Residual Approach



$\varepsilon = 10^{-3}$, AFC-SUPG Approach







¹Barrenechea, John, Knobloch: SINUM 54, 2427–2451, 2016

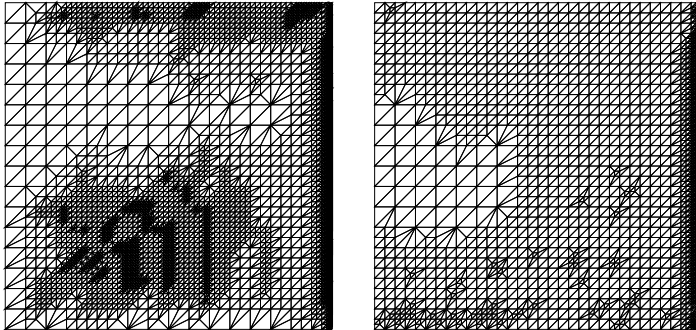


Figure 3: 14th adaptively refined grid with residual based approach. Kuzmin limiter (#nDOFs = 22962) (left) and BJK limiter (#nDOFs = 23572) (right)

- Hanging nodes
 - Preserve angles after red-refinement
 - Avoid prism and pyramids in 3D mesh refinement
- Certain stabilized schemes rely on the properties of triangulation¹

¹Xu, Zikatanov, MC, 68(228):1429-1446, 1999

²John, Knobloch, CMAME, 196(17-20):2197-2215, 2007

- Hanging nodes
 - Preserve angles after red-refinement
 - Avoid prism and pyramids in 3D mesh refinement
- Certain stabilized schemes rely on the properties of triangulation¹
- Discrete Maximum Principle

$$\text{Var}(u_h) = u_h^{\max} - u_h^{\min}$$

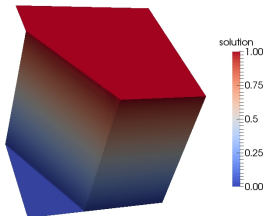
- Thickness of Interior Layer²

¹Xu, Zikatanov, MC, 68(228):1429-1446, 1999

²John, Knobloch, CMAME, 196(17-20):2197-2215, 2007

- Example with interior and boundary layer¹
- $\varepsilon = 10^{-4}$, $\mathbf{b} = (\cos(-\pi/3), \sin(-\pi/3))^T$, $\mathbf{c} = \mathbf{f} = 0$

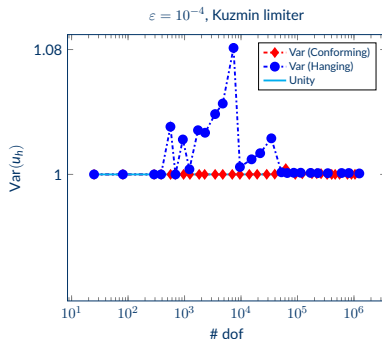
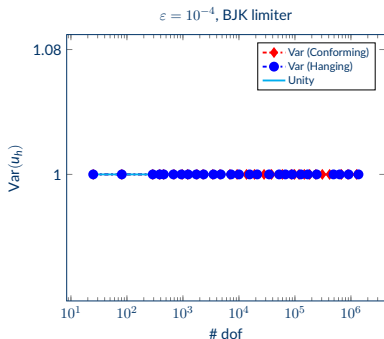
$$u_b = \begin{cases} 1 & (y = 1 \wedge x > 0) \text{ or } (x = 0 \wedge y > 0.7), \\ 0 & \text{else.} \end{cases}$$

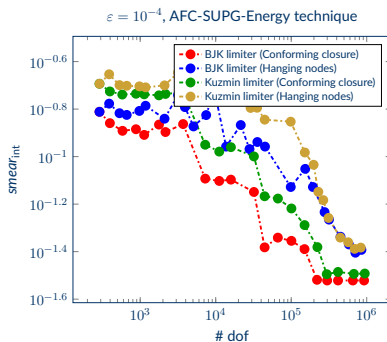
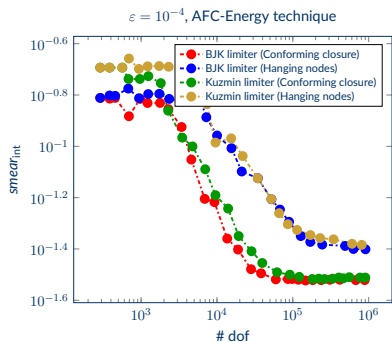


- P_1 finite elements

¹Hughes, Mallet, Mizukami, CMAME, 54(3):341-345, 1986

- $\text{Var}(u_h)$





- Iterative Solvers^{1,2}
 - With appropriate parameters some gain w.r.t. number of iterations
 - Easier to solve problems for Kuzmin limiter than for BJK limiter
 - With an appropriate solver: FPR is usually most efficient

¹ J., John: CAMWA 78, 3117-3138, 2019

² J., John: BAIL 2018, Springer 113-128, 2020

³ J.: arXiv , 2005.02938, 2020

⁴ J.: arXiv , 2007.08405 , 2020

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- A Posteriori Error Analysis³
 - Effectivity index not robust with AFC-Energy technique
 - For the AFC-SUPG-Energy technique, the effectivity index was better
 - With adaptive grid refinement, problem could become locally diffusion-dominated. Then use BJK limiter
 - For a small diffusion coefficient, use Kuzmin limiter

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- Hanging Nodes⁴
 - Kuzmin limiter fails to satisfy the DMP on grids with hanging nodes
 - BJK limiter satisfies the DMP on grids with hanging nodes
 - Layers were better resolved on grids with conforming closure
 - Numerical results on grids with hanging nodes not satisfactory

¹ J., John: CAMWA 78, 3117-3138, 2019

² J., John: BAIL 2018, Springer 113-128, 2020

³ J.: arXiv , 2005.02938, 2020

⁴ J.: arXiv , 2007.08405 , 2020

- Development of robust estimators
- Extending the analysis for local lower bound
- Stability and convergence analysis for Flux Corrected Transport (FCT) schemes for time-dependent convection-diffusion-reaction equations
- Efficient solution of the nonlinear problems in (FCT) schemes