

# Convergence analysis and numerical comparison of adaptive least-squares finite element methods

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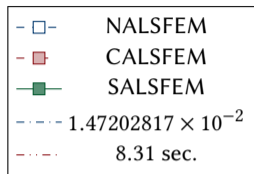
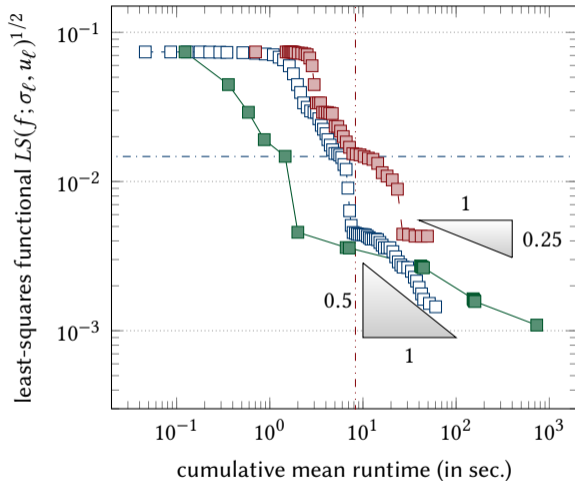


Minimum Residual & Least-Squares Finite Element Methods

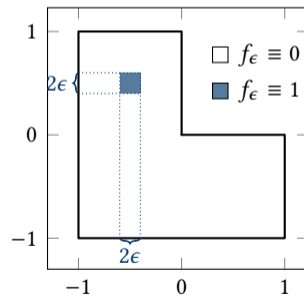
Pontificia Universidad Católica de Chile, Santiago

6<sup>th</sup> October 2022

# Motivation – Adaptive mesh-refinement



Averaged time measurements on 16 CPUs  
(Intel Xeon Platinum 8280L with max. 4.0GHz)



$$-\Delta u = f_\epsilon \text{ in } \Omega \text{ and } u = 0 \text{ on } \partial\Omega$$



**1** Adaptive least-squares FEMs

**2** Convergence analysis

**3** Numerical comparison

# Adaptive least-squares FEMs

$\Omega \subset \mathbb{R}^2$  polygonal Lipschitz domain,  $f \in L^2(\Omega)$ ,

$\partial\Omega = \Gamma_D \cup \Gamma_N$  into Dirichlet and Neumann boundary  $\Gamma_D$  and  $\Gamma_N$  with  $|\Gamma_D|, |\Gamma_N| > 0$

$$\begin{aligned} \text{Minimise } & LS(f; \sigma, u) := \|f + \operatorname{div} \sigma\|_{L^2(\Omega)}^2 + \|\sigma - \nabla u\|_{L^2(\Omega)}^2 \\ \text{s.t. } & \sigma \in \Sigma := \{\tau \in H(\operatorname{div}, \Omega) : \tau \nu = 0 \text{ on } \Gamma_N\} \\ & u \in U := \{v \in H^1(\Omega) : v = 0 \text{ on } \Gamma_D\} \end{aligned}$$

[Führer-Heuer-Karkulik, *SINUM*, 2022] for right-hand sides in  $H^{-1}(\Omega)$

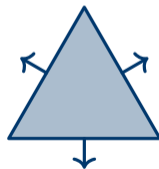
Let  $\mathcal{T}$  be regular triangulation of  $\Omega$  resolving  $\Gamma_D$  and  $\Gamma_N$  and  $k \in \mathbb{N}_0$

$$\text{Minimise } LS(f; \sigma_{LS}, u_{LS}) := \|f + \operatorname{div} \sigma_{LS}\|_{L^2(\Omega)}^2 + \|\sigma_{LS} - \nabla u_{LS}\|_{L^2(\Omega)}^2$$

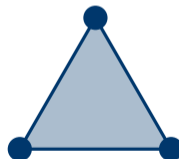
$$\text{s.t. } \sigma_{LS} \in \Sigma^k(\mathcal{T}) := \{\tau_{RT} \in RT_k(\mathcal{T}) : \tau_{RT} \nu = 0 \text{ on } \Gamma_N\}$$

$$u_{LS} \in U^{k+1}(\mathcal{T}) := \{v_C \in S^{k+1}(\mathcal{T}) : v_C = 0 \text{ on } \Gamma_D\}$$

$k = 0$



$RT_0$



$S^1$

For subsets  $\mathcal{M} \subseteq \mathcal{T}$  define the error estimator  $\eta_N^2(\mathcal{T}, \mathcal{M}) := \sum_{T \in \mathcal{M}} \eta_N^2(\mathcal{T}, T)$  with

$$\eta_N^2(\mathcal{T}, T) := \|f + \operatorname{div} \sigma_{\text{LS}}\|_{L^2(T)}^2 + \|\sigma_{\text{LS}} - \nabla u_{\text{LS}}\|_{L^2(T)}^2$$

Reliability and efficiency

$$\eta_N^2(\mathcal{T}) := \eta_N^2(\mathcal{T}, \mathcal{T}) = LS(f; \sigma_{\text{LS}}, u_{\text{LS}}) \approx \|\sigma - \sigma_{\text{LS}}\|_{H(\operatorname{div}, \Omega)}^2 + \|\nabla(u - u_{\text{LS}})\|_{L^2(\Omega)}^2$$

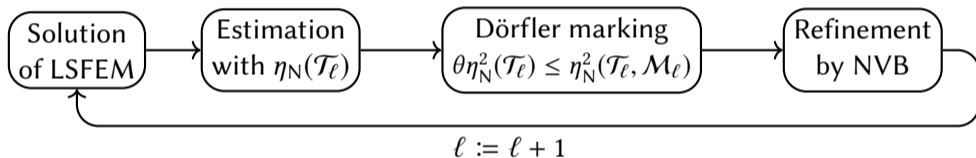
Asymptotic exactness

$$LS(f; \sigma_{\text{LS}}, u_{\text{LS}}) / (\|\sigma - \sigma_{\text{LS}}\|_{H(\operatorname{div}, \Omega)}^2 + \|\nabla(u - u_{\text{LS}})\|_{L^2(\Omega)}^2) \rightarrow 1 \quad \text{as} \quad h_{\max} \rightarrow 0$$

[Carstensen-Storn, *SINUM*, 2018]

## NALSFEM

Input: initial triangulation  $\mathcal{T}_0$  and bulk parameter  $0 < \theta \leq 1$



Output: triangulations  $\mathcal{T}_\ell$  with solutions  $\sigma_\ell, u_\ell$

[Dörfler, *SINUM*, 1996] [Pfeiler-Praetorius, *Math. Comp.* 2020] [Stevenson, *Math. Comp.* 2008]



Let  $\eta_S^2(\mathcal{T}) := \sum_{T \in \mathcal{T}} \eta_S^2(\mathcal{T}, T)$ ,  $\mu^2(\mathcal{T}) := \sum_{T \in \mathcal{T}} \mu^2(T)$ , and  $\text{osc}^2(f, \mathcal{T}) := \sum_{T \in \mathcal{T}} \text{osc}^2(f, T)$  with

$$\begin{aligned} \eta_S^2(\mathcal{T}, T) &:= |T| \left( \|\text{div}(\sigma_{\text{LS}} - \nabla u_{\text{LS}})\|_{L^2(T)}^2 + \|\text{curl } \sigma_{\text{LS}}\|_{L^2(T)}^2 \right) \\ &\quad + |T|^{1/2} \sum_{F \in \mathcal{F}(T) \setminus \mathcal{F}(\Gamma_D)} \|[\nabla u_{\text{LS}} \cdot n_F]_F\|_{L^2(F)}^2 \\ &\quad + |T|^{1/2} \sum_{F \in \mathcal{F}(T) \setminus \mathcal{F}(\Gamma_N)} \|[\sigma_{\text{LS}} \cdot t_F]_F\|_{L^2(F)}^2 \end{aligned}$$

$$\mu^2(T) := \|(1 - \Pi_k)f\|_{L^2(T)}^2$$

$$\text{osc}^2(f, T) := |T|^{1/2} \|(1 - \Pi_k)f\|_{L^2(T)}^2$$

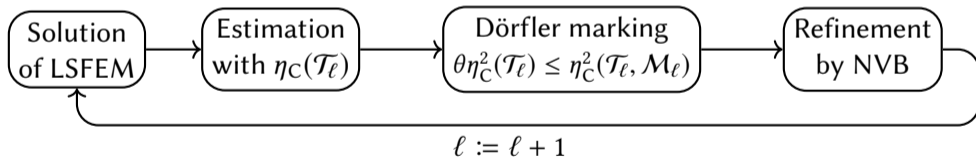
[Carstensen-Park, *SINUM*, 2015]

The estimator  $\eta_C^2(\mathcal{T}, T) := \eta_S^2(\mathcal{T}, T) + \text{osc}^2(f, T)$  satisfies

$$\eta_C^2(\mathcal{T}) := \sum_{T \in \mathcal{T}} \eta_C^2(\mathcal{T}, T) \approx \|\sigma - \sigma_{\text{LS}}\|_{L^2(\Omega)}^2 + \|\nabla(u - u_{\text{LS}})\|_{L^2(\Omega)}^2 + \text{osc}^2(f, \mathcal{T})$$

## CALSFEM

**Input:** initial triangulation  $\mathcal{T}_0$  and bulk parameter  $0 < \theta \leq 1$



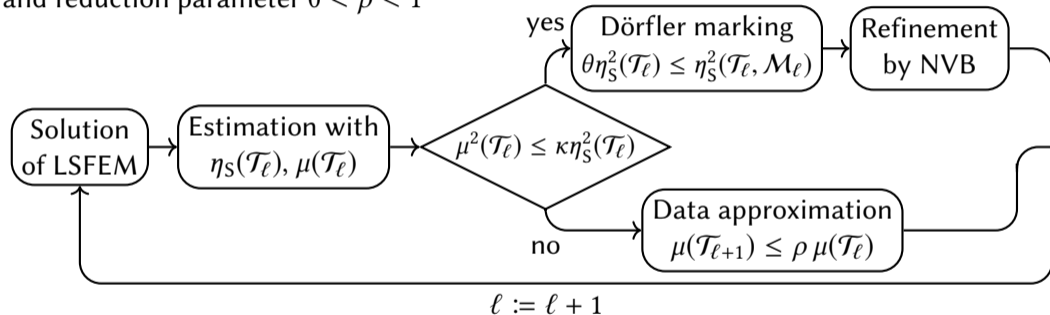
**Output:** triangulations  $\mathcal{T}_\ell$  with solutions  $\sigma_\ell, u_\ell$

[Carstensen, *Math. Comp.* 2020]

[Carstensen-Ma, *CAMWA*, 2021]

## SALSFEM

**Input:** initial triangulation  $\mathcal{T}_0$ , separation parameter  $0 < \kappa$ , bulk parameter  $0 < \theta \leq 1$ , and reduction parameter  $0 < \rho < 1$



**Output:** triangulations  $\mathcal{T}_\ell$  with solutions  $\sigma_\ell, u_\ell$

[Carstensen-Park, *SINUM*, 2015]

[Binev-DeVore, *Numer. Math.* 2004]  
[Carstensen-Rabus, *Math. Comp.* 2011]

## Convergence analysis



The output  $(\sigma_\ell, u_\ell)_\ell$  of NALSFEM satisfies

$$\|\sigma - \sigma_\ell\|_{H(\operatorname{div}, \Omega)}^2 + \|\nabla(u - u_\ell)\|_{L^2(\Omega)}^2 + \eta_N^2(\mathcal{T}_\ell) \rightarrow 0 \quad \text{as } \ell \rightarrow \infty$$

even for other marking strategies

[Führer-Praetorius, *CAMWA*, 2020; Gantner-Stevenson, *M2AN*, 2021; Siebert, *IMA JNA*, 2011]



For  $k = 0$

Assume that the initial triangulation is sufficiently fine in that  $f = \Pi_{L+1}f$

There exist a **minimal bulk parameter**  $0 < \theta_0 < 1$  and constants  $0 < \rho < 1$  and  $0 < \Lambda < \infty$  such that for all  $\theta_0 \leq \theta \leq 1$ , the modified estimator

$$\tilde{\eta}_N^2(\mathcal{T}_\ell) := LS(f; \sigma_\ell, u_\ell) + \Lambda \|(1 - \Pi_\ell)\sigma_\ell\|_{L^2(\Omega)}^2$$

satisfies

$$\tilde{\eta}_N^2(\mathcal{T}_{\ell+1}) \leq \rho \tilde{\eta}_N^2(\mathcal{T}_\ell) \quad \text{for all } \ell = L, L+1, \dots$$

[Carstensen-Park-Bringmann, *Numer. Math.* 2017]



There exists a **maximal bulk parameter**  $0 < \theta_0 < 1$  such that for all  $0 < \theta < \theta_0$  and  $0 < s < \infty$ , the output  $(\sigma_\ell, u_\ell)_\ell$  of CALSFEM satisfies

$$\sup_{N \in \mathbb{N}} (N + 1)^s \min_{\mathcal{T} \in \mathbb{T}(N)} \eta_C(\mathcal{T}) \approx \sup_{\ell \in \mathbb{N}} (|\mathcal{T}_\ell| - |\mathcal{T}_0| + 1)^s \eta_C(\mathcal{T}_\ell)$$

[Carstensen, *Math. Comp.* 2020]

[Carstensen-Ma, *CAMWA*, 2021]

**Proof** by axioms of adaptivity

[Carstensen et al. *CAMWA*, 2014]



There exist a **maximal bulk parameter**  $0 < \theta_0 < 1$  and a **maximal separation parameter**  $0 < \kappa_0 \leq \infty$  such that for all  $0 < \theta < \theta_0$ ,  $0 < \kappa < \kappa_0$ ,  $0 < \rho < 1$ , and  $0 < s < \infty$ , the output  $(\sigma_\ell, u_\ell)_\ell$  of SALSFEM satisfies

$$\sup_{N \in \mathbb{N}} (N + 1)^s \min_{\mathcal{T} \in \mathbb{T}(N)} (\eta_S^2(\mathcal{T}) + \mu^2(\mathcal{T}))^{1/2} \approx \sup_{\ell \in \mathbb{N}} (|\mathcal{T}_\ell| - |\mathcal{T}_0| + 1)^s (\eta_S^2(\mathcal{T}_\ell) + \mu^2(\mathcal{T}_\ell))^{1/2}$$

**Proof** by axioms of adaptivity

[Carstensen-Rabus, *SINUM*, 2017]

**Unified proof** for Poisson problem, Stokes equations, and linear elasticity in 2D and 3D with inhomogeneous boundary conditions and arbitrary polynomial degree  $k \in \mathbb{N}_0$  presented in

[Bringmann, *PhD thesis*, 2021]

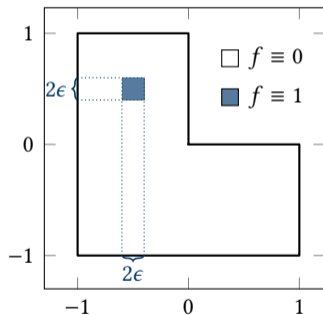
[Bringmann, *J. Numer. Math.* 2022]



## Numerical comparison

Poisson model problem with homogeneous Dirichlet boundary conditions

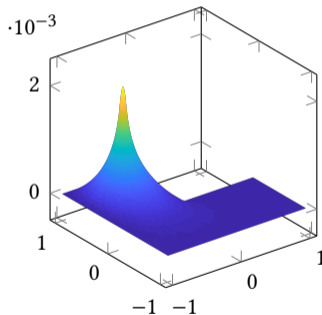
$$f_\epsilon \in L^2(\Omega) \quad \text{with} \quad \epsilon > 0 \quad \text{and} \quad f_\epsilon(x) := \begin{cases} 1 & \text{if } |x_1 + \frac{1}{2}| \leq \epsilon \text{ and } |x_2 - \frac{1}{2}| \leq \epsilon \\ 0 & \text{otherwise} \end{cases}$$



L-shaped domain  $\Omega = (-1, 1)^2 \setminus [0, 1)^2$

[Rabus, *J. Numer. Math.* 2015, Sect. 3.4]

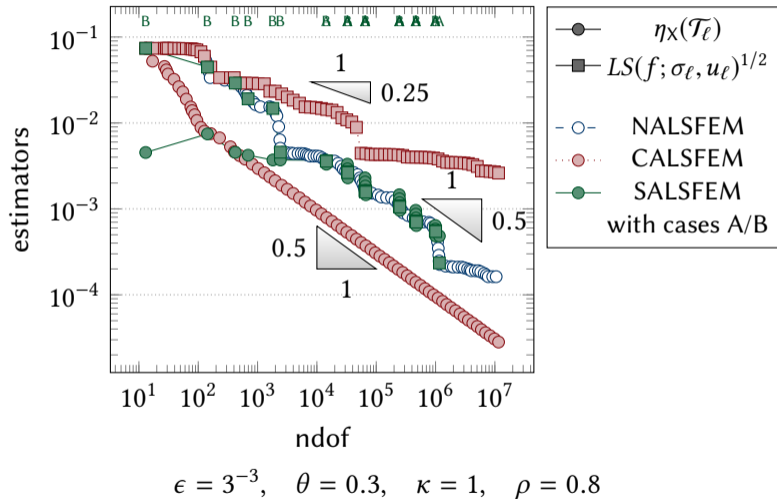
P. Bringmann



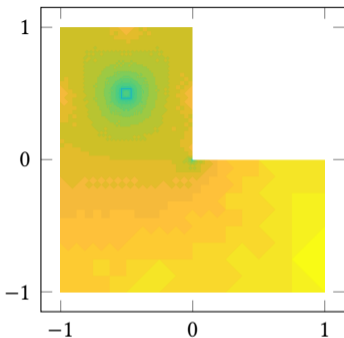
Approximation to solution of  $-\Delta u = f_\epsilon$

[Sutherland-Hodgman, *Commun. ACM*, 1974]

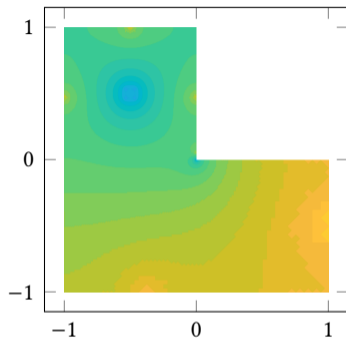
# Comparison of the three methods



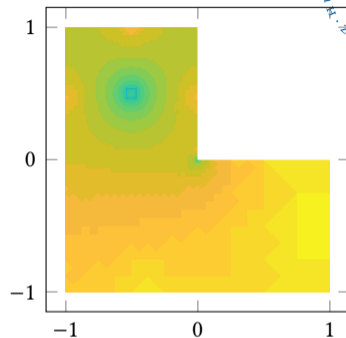
# Comparison of meshes



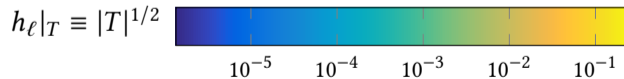
NALSFEM (575 130 triangles)



CALSFEM (551 344 triangles)

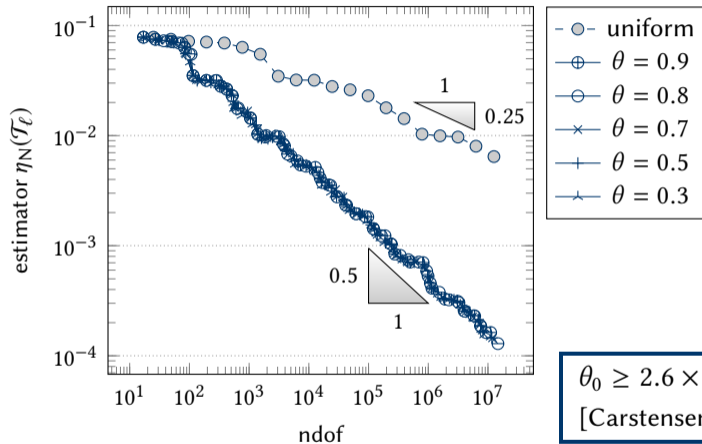


SALSFEM (580 208 triangles)



$$\epsilon = 3^{-3}, \quad \theta = 0.3, \quad \kappa = 1, \quad \rho = 0.8$$

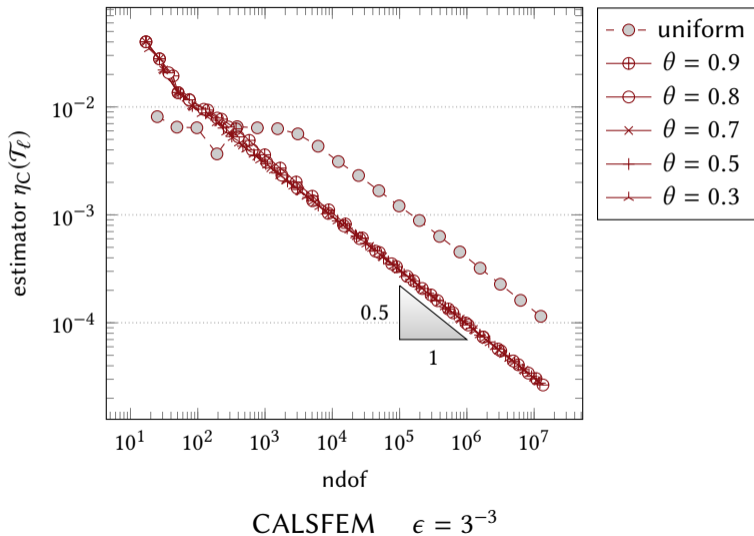
# Choice of bulk parameter $\theta$



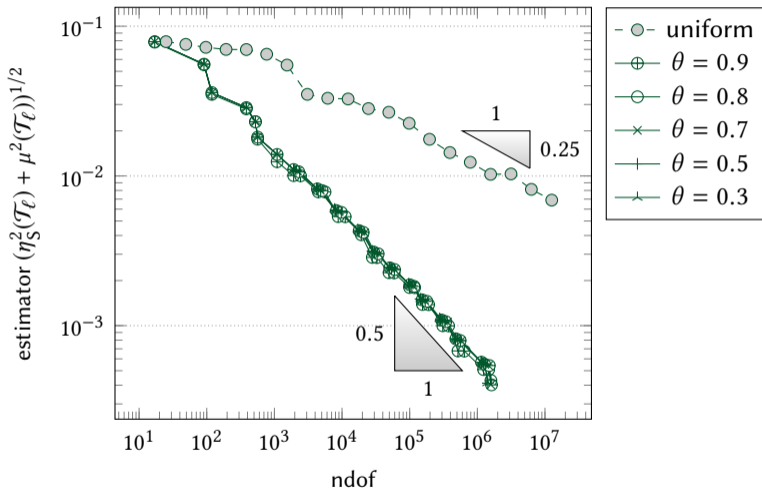
NALSFEM  $\epsilon = 3^{-3}$

$\theta_0 \geq 2.6 \times 10^{-6}$  for Courant FEM  
[Carstensen-Hellwig, *CMAM*, 2018]

# Choice of bulk parameter $\theta$

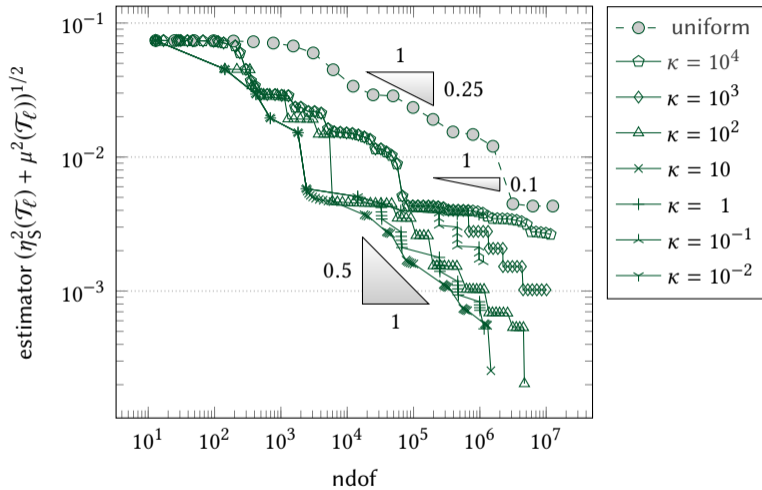


# Choice of bulk parameter $\theta$



SALSFEM  $\epsilon = 3^{-3}$ ,  $\kappa = 10$ ,  $\rho = 0.8$

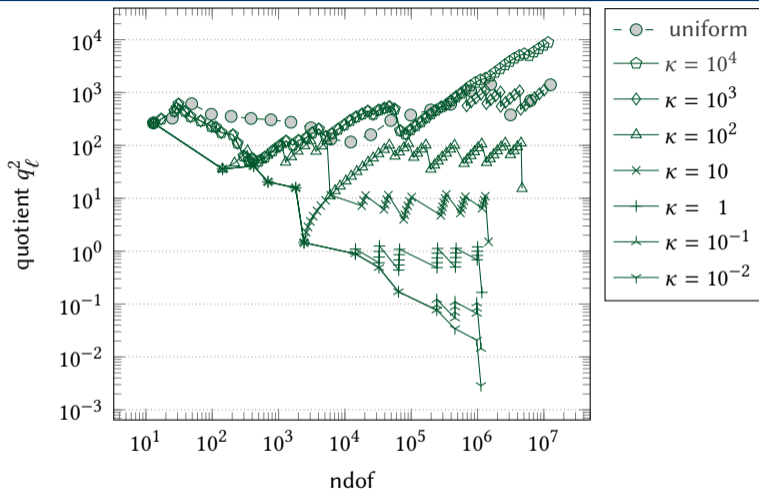
# Choice of separation parameter $\kappa$ in SALSFEM



$$\epsilon = 3^{-3}, \quad \theta = 0.3, \quad \rho = 0.8$$

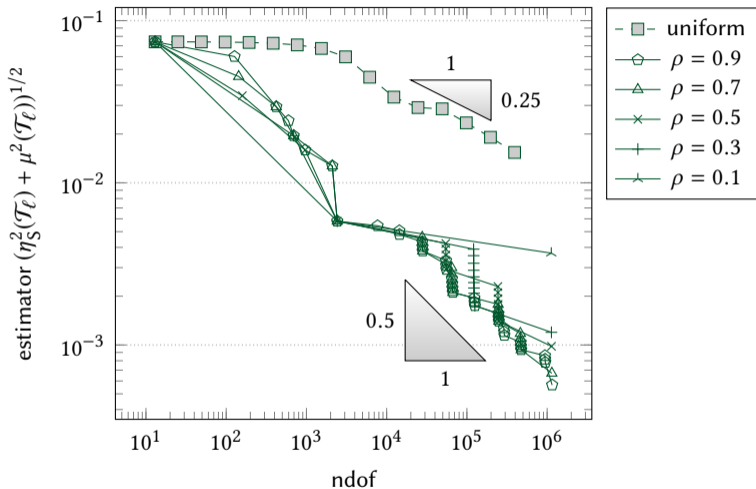


# Choice of separation parameter $\kappa$ in SALSFEM



$$q_\ell^2 := \mu^2(\mathcal{T}_\ell) / \eta_S^2(\mathcal{T}_\ell)$$

$$\epsilon = 3^{-3}, \quad \theta = 0.3, \quad \rho = 0.8$$



$$\epsilon = 3^{-3}, \quad \theta = 0.3, \quad \kappa = 1$$

## Choice of method

SALSFEM for guaranteed optimal convergence rates

NALSFEM for efficient alternative

CALSFEM for efficient method with optimal convergence rates in weaker norm

## Choice of parameters

bulk parameter  $0.3 \leq \theta \leq 0.7$

separation parameter  $\kappa := q_\ell^2/10$  for  $\ell = 0$  or small  $\ell > 0$  (e.g., from uniform refinement)

reduction parameter  $0.3 \leq \rho \leq 0.8$

# Thank you for your attention!



The speaker acknowledges the support from



## Publications

P. Bringmann. *Adaptive least-squares finite element method with optimal convergence rates*. *PhD thesis* 2021. Humboldt-Universität zu Berlin.

P. Bringmann. *Computational competition of three adaptive least-squares finite element schemes*. Submitted. Preprint available at [arXiv:2209.06028](https://arxiv.org/abs/2209.06028). 2022.

P. Bringmann. *How to prove optimal convergence rates for adaptive least-squares finite element methods*. *J. Numer. Math.* 2022. In press and published online. DOI: [10.1515/jnma-2021-0116](https://doi.org/10.1515/jnma-2021-0116).

P. Bringmann, C. Carstensen, and N. T. Tran. 'Adaptive least-squares, discontinuous Petrov-Galerkin, and hybrid high-order methods'. *Non-standard discretisation methods in solid mechanics*. Vol. 98. Lect. Notes Appl. Comput. Mech. Final survey of projects CA 151/22-1&2 in SPP 1748. Springer, Cham, 2022.

P. Bringmann



[Carstensen-Park, *SINUM*, 2015]

Poisson model problem in 2D, homogeneous Dirichlet BC, lowest-order discretisation

[Bringmann-Carstensen, *Numer. Math.* 2017]

Stokes equations in 2D, **inhomogeneous Dirichlet BC**, lowest-order discretisation

[Bringmann-Carstensen, *CAMWA*, 2017]

Stokes equations in 2D, inhomogeneous Dirichlet BC, **higher-order discretisation**

[Bringmann-Carstensen-Starke, *SINUM*, 2018]

Linear elasticity in **3D**, mixed **inhomogeneous Neumann BC**, lowest-order discretisation

[Bringmann, *PhD thesis*, 2021]

Generalised model problem in 3D, mixed inhomogeneous BC, higher-order discretisation



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