

Preconditioned iterative solvers for immersed finite element methods

Frits de Prenter,
Clemens Verhoosel & Harald van Brummelen

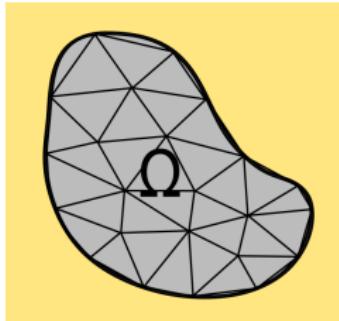
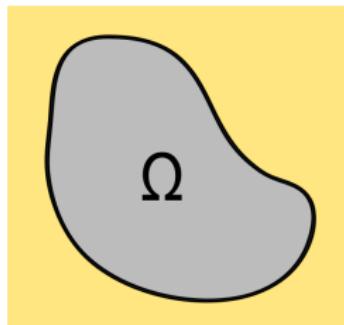
Eindhoven University of Technology
Department of Mechanical Engineering
Energy Technology Fluid Dynamics

May 19, 2017

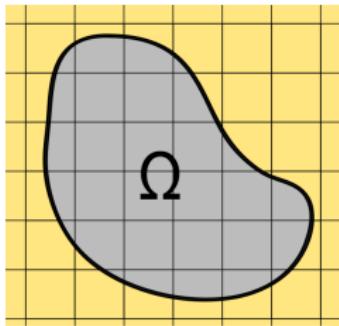
Outline

- 1** Immersed finite element methods
- 2** Conditioning analysis and preconditioning

Concept



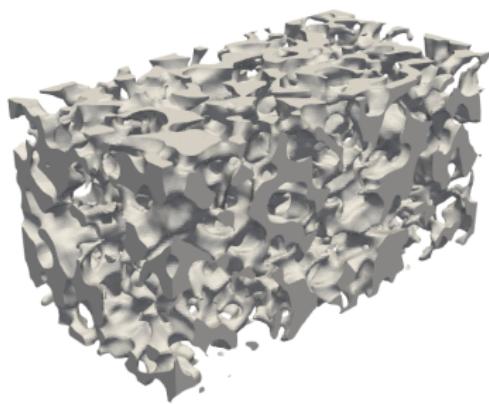
Conforming FEM



Immersed FEM

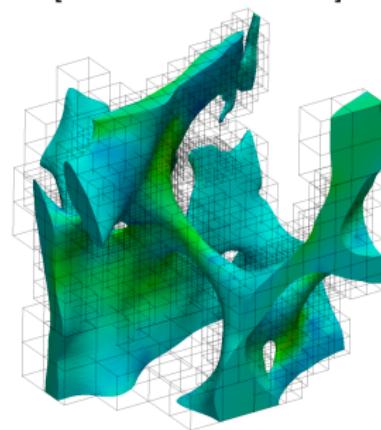
Motivation (1): complex geometries

[C.-Z. Qin]



Sintered glass beads

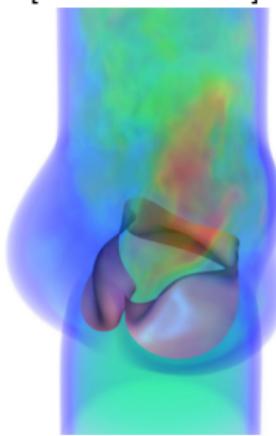
[C.V. Verhoosel 2015]



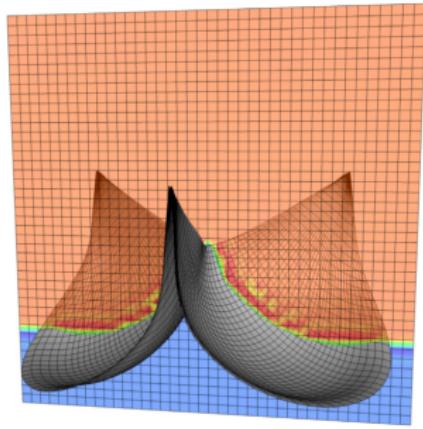
Trabecular bone

Motivation (2): time dependent domains

[M.-C. Hsu 2014]



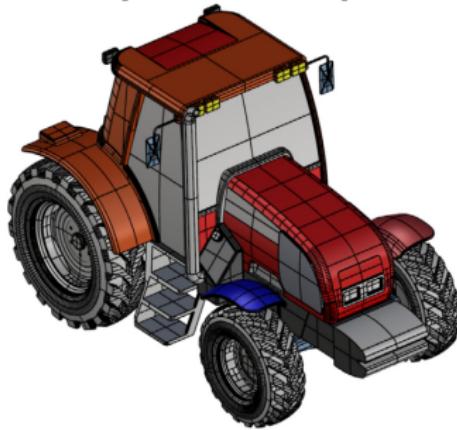
[D. Kamensky 2017]



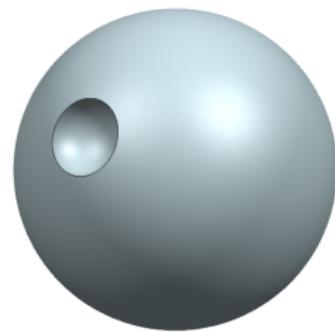
Transient simulation of prosthetic heart valve

Motivation (3): isogeometric analysis (IGA)

[M.-C. Hsu 2016]

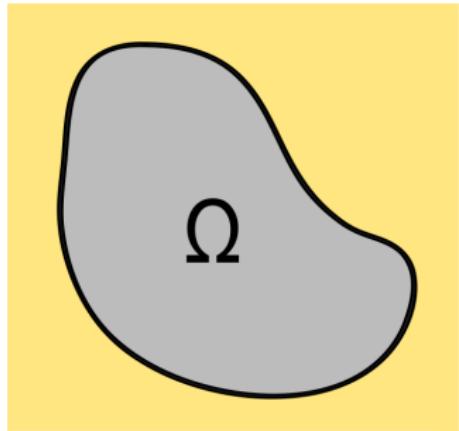


External flow around
CAD-geometry



IGA on trimmed CAD-geometries

Imposing boundary conditions



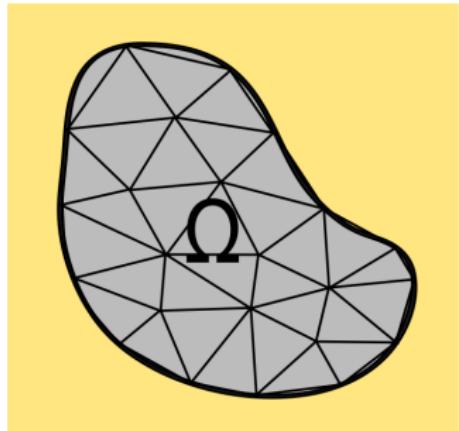
Domain

$$\begin{aligned} \text{strong} & \left\{ \begin{array}{l} -\Delta u = f \text{ in } \Omega \\ u = g^D \text{ on } \Gamma^D \subset \partial\Omega \\ n \cdot \nabla u = g^N \text{ on } \Gamma^N \subset \partial\Omega \end{array} \right. \\ \text{weak} & \left\{ \begin{array}{l} \text{find } w \in \mathcal{H}_0^1(\Omega) \text{ s.t. :} \\ a(v, w) = b(v) - a(v, q) \\ \text{for all } v \in \mathcal{H}_0^1(\Omega) \end{array} \right. \end{aligned}$$

$$a(v, u) = \int_{\Omega} \nabla v \cdot \nabla u \, dV$$

$$b(v) = \int_{\Omega} vf \, dV + \int_{\Gamma^N} vg^N \, dS$$

Imposing boundary conditions



strong $\begin{cases} -\Delta u &= f \text{ in } \Omega \\ u &= g^D \text{ on } \Gamma^D \subset \partial\Omega \\ n \cdot \nabla u &= g^N \text{ on } \Gamma^N \subset \partial\Omega \end{cases}$

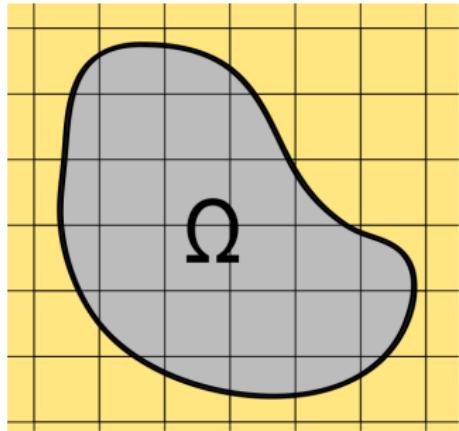
weak $\begin{cases} \text{find } w^h \in \mathcal{V}_0^h(\Omega) \subset \mathcal{H}_0^1(\Omega) \text{ s.t. :} \\ a(v, w) = b(v) - a(v, q) \\ \text{for all } v^h \in \mathcal{V}_0^h(\Omega) \subset \mathcal{H}_0^1(\Omega) \end{cases}$

Conforming FEM

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Imposing boundary conditions



Immersed FEM

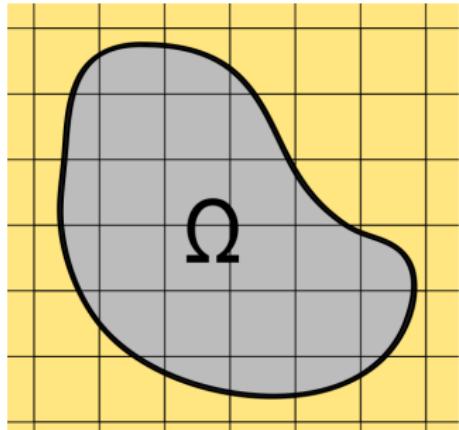
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Imposing boundary conditions



Immersed FEM

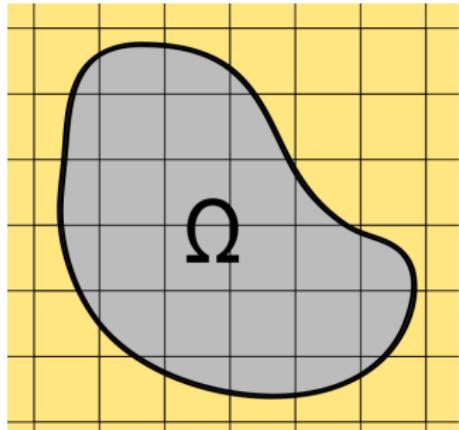
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Imposing boundary conditions



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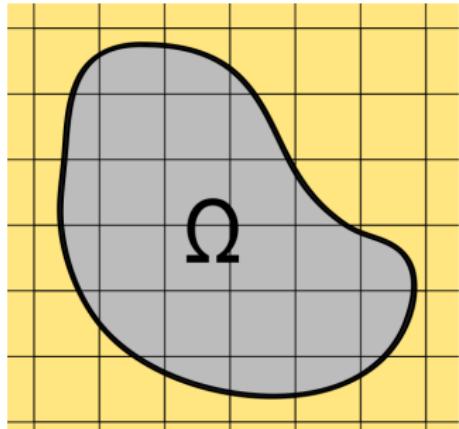
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Immersed FEM

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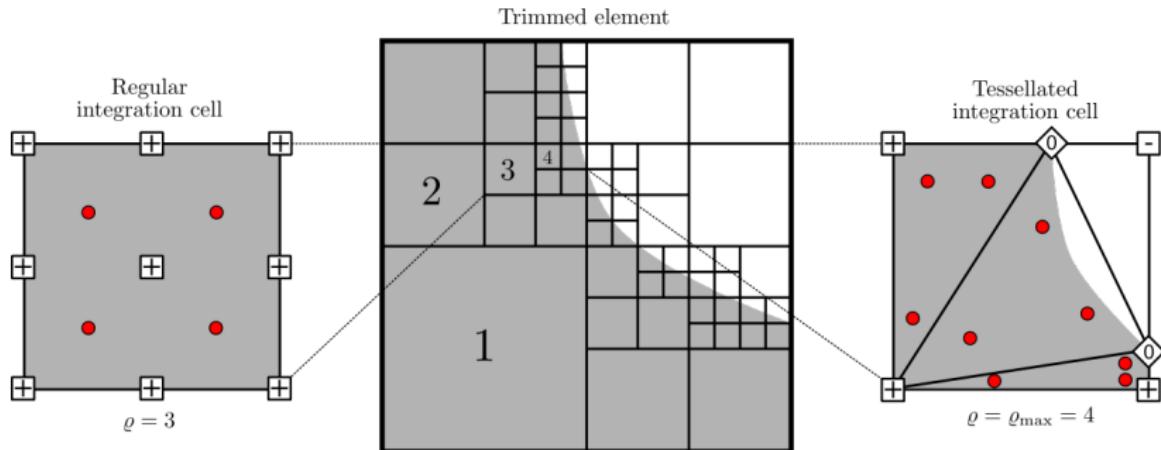
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Immersed FEM

$$a(v, u) = \int_{\Omega} \nabla v \cdot \nabla u dV + \int_{\Gamma^D} -v(n \cdot \nabla u) - (n \cdot \nabla v)u + \beta vu dS$$

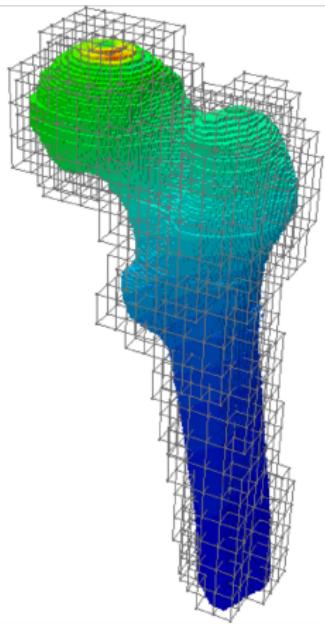
$$b(v) = \int_{\Omega} vf dV + \int_{\Gamma^N} vg^N dS + \int_{\Gamma^D} -(n \cdot \nabla v)g^D + \beta vg^D dS$$

Integration of trimmed elements

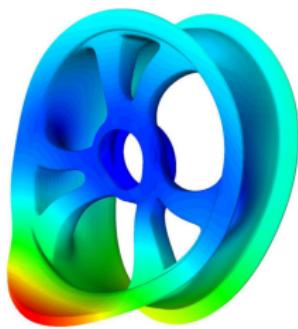


[C.V. Verhoosel 2015]

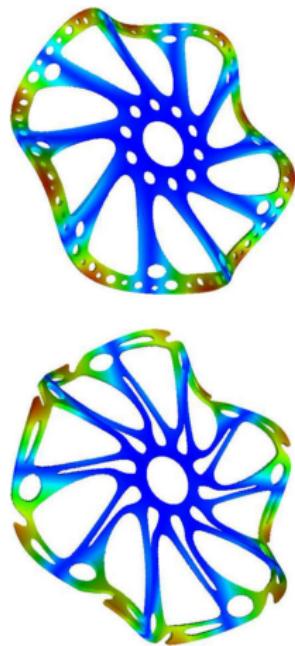
Static elasticity problems



[Ruess 2013]

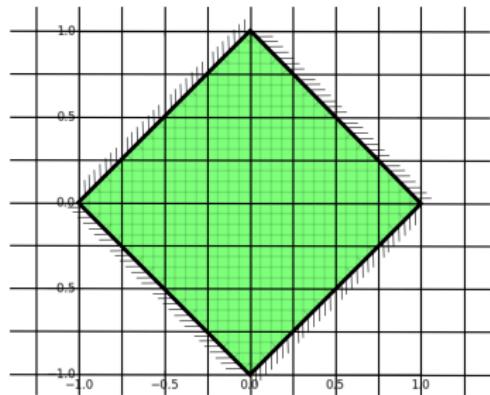


[Schillinger 2012]

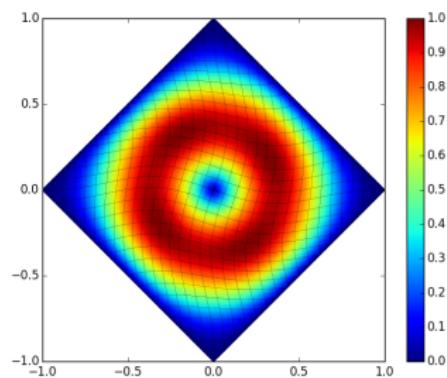


[Rank 2012]

Dynamic elasticity problems

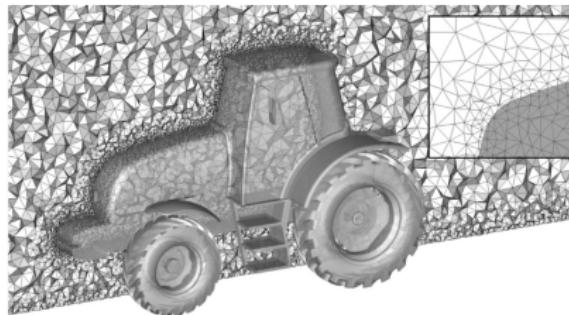


Grid



Initial condition

Large Eddy Simulations (LES)



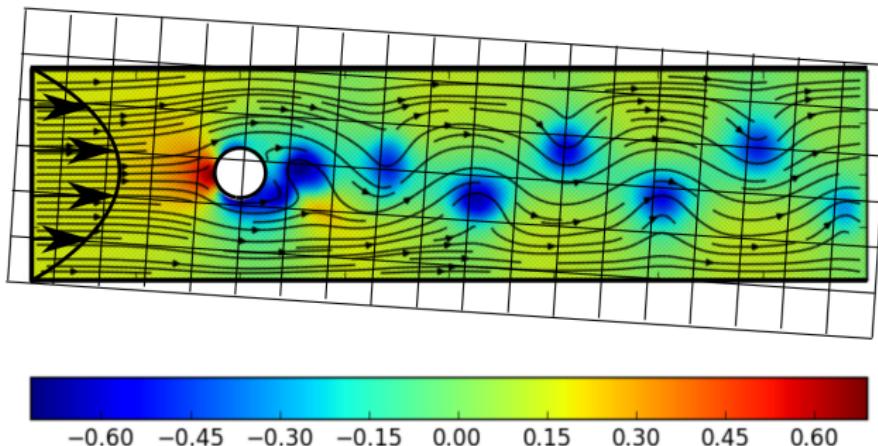
[M.-C. Hsu 2016]



[F. Xu 2016]

Variational Multiscale modeling of Navier-Stokes

Transient flow problems



Von Karman vortex street in Navier-Stokes

Outline

- 1** Immersed finite element methods
- 2** Conditioning analysis and preconditioning

Conditioning of immersed methods

From weak form to linear system

$$\text{function} \quad v^h \left(= \Phi^T \mathbf{v} \right) \Leftrightarrow \mathbf{v} \quad \text{coefficient vector}$$

$$\text{weak form} \quad a(v^h, u^h) = b(v^h) \Leftrightarrow \mathbf{A}\mathbf{u} = \mathbf{b} \quad \text{linear system}$$

$$\text{condition number: } \kappa(\mathbf{A}) = \|\mathbf{A}\| \|\mathbf{A}^{-1}\|$$

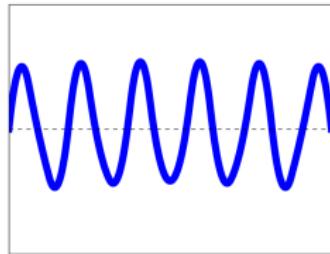
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v^h : highest periodicity

$$\|\mathbf{A}\| = \max_{\mathbf{v} \neq 0} \frac{\|\mathbf{A}\mathbf{v}\|}{\|\mathbf{v}\|} = \max_{\|\mathbf{v}\|=1} \left\| a \left(\Phi, v^h \right) \right\|$$

SPD systems :

$$\max_{\|\mathbf{v}\|=1} \mathbf{v}^T \mathbf{A} \mathbf{v} = \max_{\|\mathbf{v}\|=1} a(v^h, v^h)$$

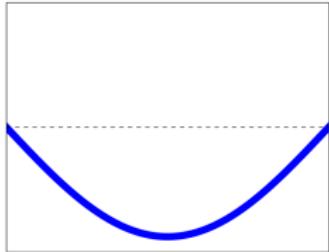
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v^h : lowest periodicity

$$\|\mathbf{A}^{-1}\| = \max_{\mathbf{v} \neq 0} \frac{\|\mathbf{v}\|}{\|\mathbf{A}\mathbf{v}\|} = \max_{\|\mathbf{v}\|=1} \frac{1}{\|a(\Phi, v^h)\|}$$

SPD systems :

$$\max_{\|\mathbf{v}\|=1} \frac{1}{\mathbf{v}^T \mathbf{A} \mathbf{v}} = \max_{\|\mathbf{v}\|=1} \frac{1}{a(v^h, v^h)}$$

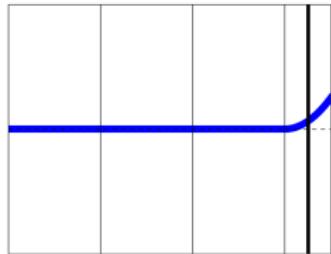
Conditioning of immersed methods

From weak form to linear system

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v^h : lowest periodicity

v^h : smallest volume fraction!

$$\|\mathbf{A}^{-1}\| = \max_{\mathbf{v} \neq 0} \frac{\|\mathbf{v}\|}{\|\mathbf{A}\mathbf{v}\|} = \max_{\|\mathbf{v}\|=1} \frac{1}{\|a(\Phi, v^h)\|}$$

$$\eta = \min_e \frac{|\Omega_{\text{cut}}^e|}{|\Omega_{\text{uncut}}^e|}$$

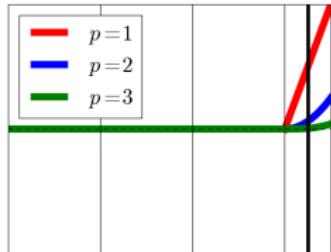
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order dependence!

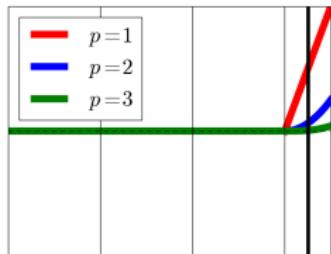
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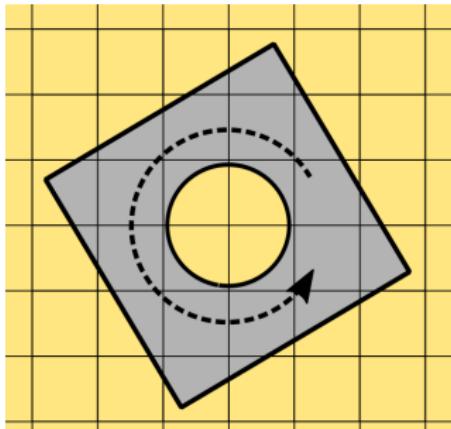
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$$\kappa \propto \eta^{-(2p+n)}$$

Verification of conditioning analysis

Domain:

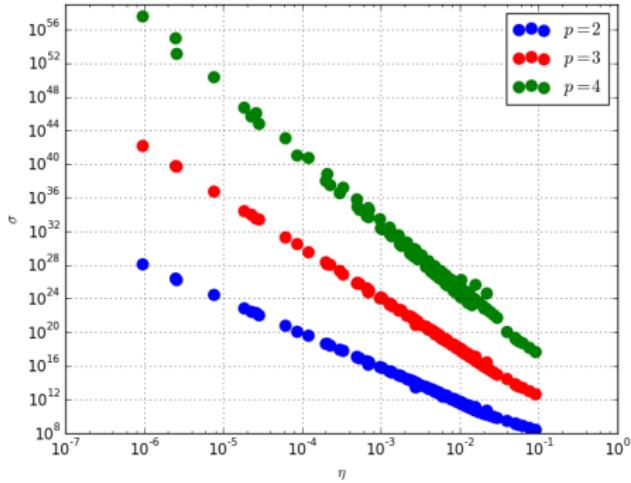


Experiment

- Domain rotated over grid
- *Different discretizations of the same problem with the same mesh size*
- κ (condition number) and η (volume fraction) at every separate rotation

Verification of conditioning analysis

Stokes:

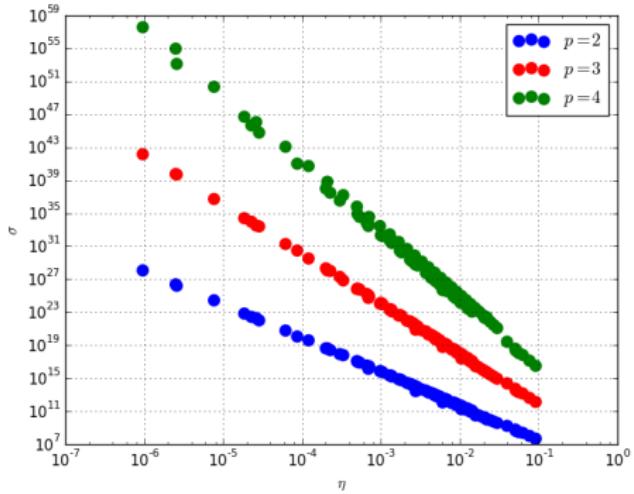


Experiment

- Domain rotated over grid
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Verification of conditioning analysis

Navier-Stokes:



Experiment

- Domain rotated over grid
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Preconditioning concept

Problem analysis

Functions v^h and corresponding coefficient vectors \mathbf{v} with:

$$v^h \ll \mathbf{v} \quad (1)$$

Preconditioning the space

- Replace basis Φ by the manipulated basis $\bar{\Phi} = \mathbf{S}\Phi$
- For nonsingular \mathbf{S} the bases Φ and $\bar{\Phi}$ span the same space
- Choose matrix \mathbf{S} such that the problem in (1) is precluded

Implementation

The preconditioned system becomes:

$$\mathbf{S}\mathbf{A}\mathbf{S}^T \bar{\mathbf{u}} = \mathbf{S}\mathbf{b}, \quad \mathbf{u} = \mathbf{S}^T \bar{\mathbf{u}}$$

This has the same eigenvalues as the left preconditioned system:

$$\mathbf{S}^T \mathbf{S} \mathbf{A} \mathbf{u} = \mathbf{S}^T \mathbf{S} \mathbf{b}$$

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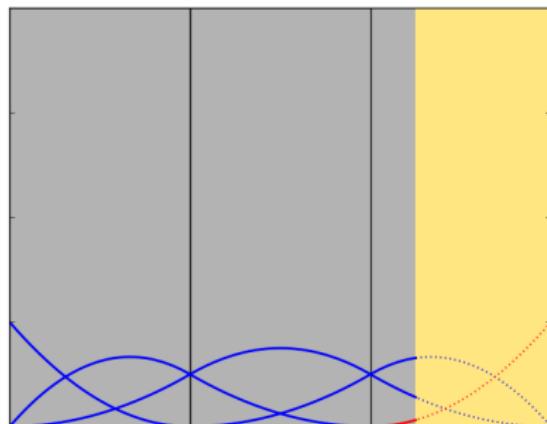
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What \mathbf{S} does (1): Scaling



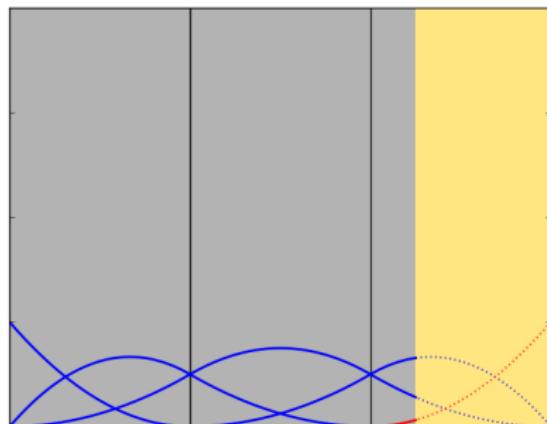
Original basis Φ

Small basis functions

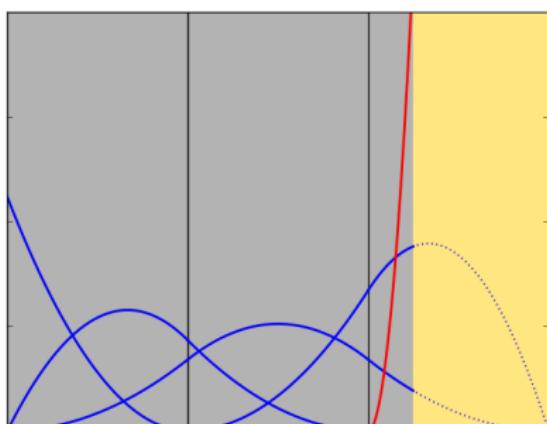
If a basis function ϕ is small, then the (unit) vector $\|\mathbf{w}\| = 1$ corresponding to $w^h = \phi$ yields $\|\mathbf{Aw}\| \ll 1$

$$\|\mathbf{A}^{-1}\| = \max_{\mathbf{v} \neq 0} \frac{\|\mathbf{v}\|}{\|\mathbf{Av}\|} \geq \frac{\|\mathbf{w}\|}{\|\mathbf{Aw}\|} \gg 1$$

What \mathbf{S} does (1): Scaling



Original basis Φ



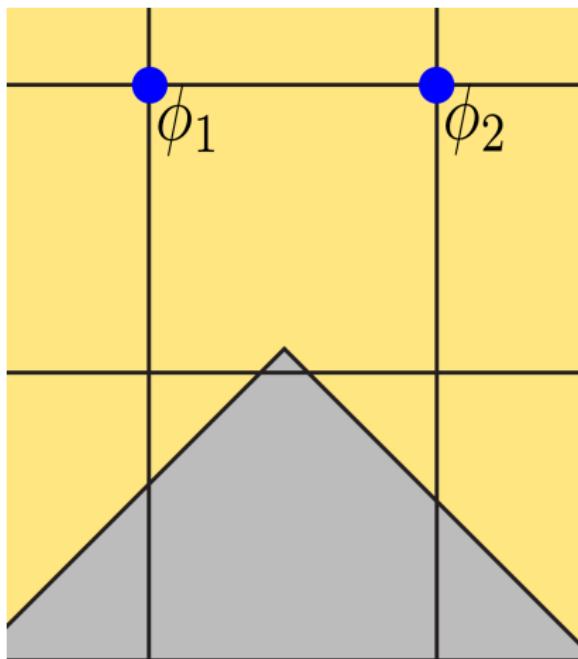
Scaled basis $\tilde{\Phi} = \mathbf{D}\Phi$

Small basis functions

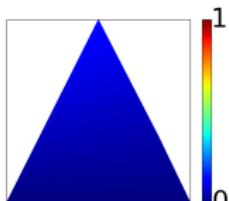
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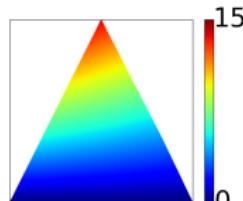
What \mathbf{S} does (2): Local orthonormalization



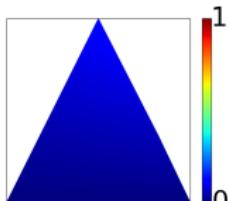
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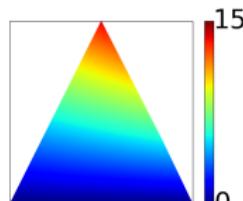
ϕ_1



$\tilde{\phi}_1 = \phi_1 / \|\phi_1\|$



ϕ_2



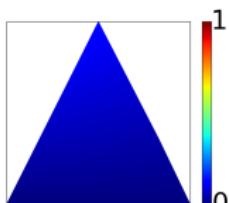
$\tilde{\phi}_2 = \phi_2 / \|\phi_2\|$

Quasi linear dependencies

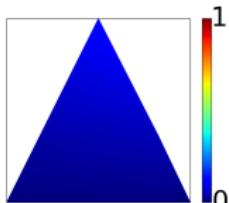
If basis functions $\tilde{\phi}_1$ and $\tilde{\phi}_2$ are very similar, then the vector $\|w\| = \sqrt{2}$ corresponding to $w^h = \tilde{\phi}_1 - \tilde{\phi}_2$ yields $\|DADw\| \ll 1$

$$\|DAD^{-1}\| = \max_{v \neq 0} \frac{\|v\|}{\|DADv\|} \geq \frac{\|w\|}{\|DADw\|} \gg 1$$

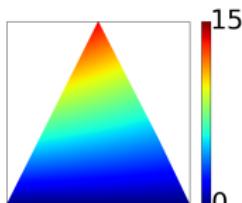
What \mathbf{S} does (2): Local orthonormalization



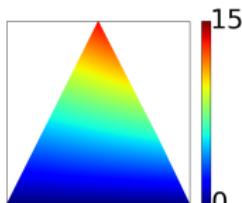
ϕ_1



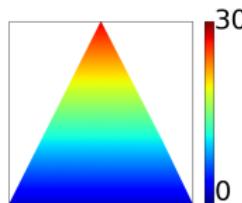
ϕ_2



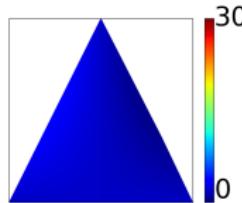
$\tilde{\phi}_1 = \phi_1 / \|\phi_1\|$



$\tilde{\phi}_2 = \phi_2 / \|\phi_2\|$



$\tilde{\phi}_1 + \tilde{\phi}_2$



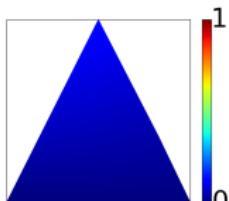
$\tilde{\phi}_1 - \tilde{\phi}_2$

Quasi linear dependencies

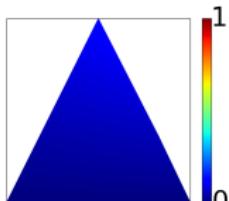
If basis functions $\tilde{\phi}_1$ and $\tilde{\phi}_2$ are very similar, then the vector $\|\mathbf{w}\| = \sqrt{2}$ corresponding to $w^h = \tilde{\phi}_1 - \tilde{\phi}_2$ yields $\|\mathbf{DADw}\| \ll 1$

$$\|\mathbf{DAD}^{-1}\| = \max_{\mathbf{v} \neq 0} \frac{\|\mathbf{v}\|}{\|\mathbf{DADv}\|} \geq \frac{\|\mathbf{w}\|}{\|\mathbf{DADw}\|} \gg 1$$

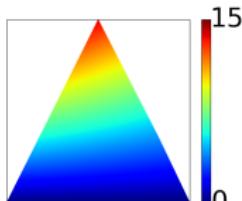
What \mathbf{S} does (2): Local orthonormalization



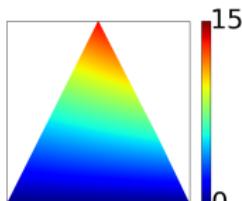
ϕ_1



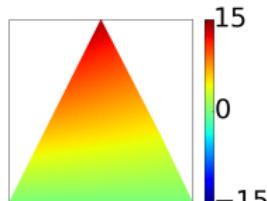
ϕ_2



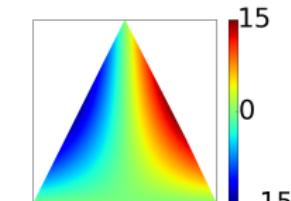
$\tilde{\phi}_1 = \phi_1 / \|\phi_1\|$



$\tilde{\phi}_2 = \phi_2 / \|\phi_2\|$



$\bar{\phi}_1 = \tilde{\phi}_1$



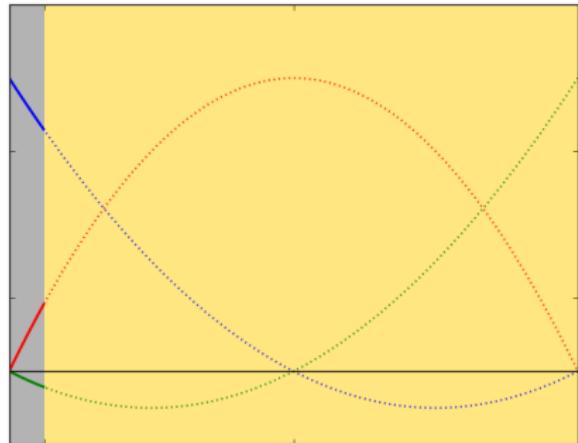
$\bar{\phi}_2 \propto \tilde{\phi}_2 + \alpha \tilde{\phi}_1$

Quasi linear dependencies

If basis functions $\tilde{\phi}_1$ and $\tilde{\phi}_2$ are very similar, then the vector $\|\mathbf{w}\| = \sqrt{2}$ corresponding to $w^h = \tilde{\phi}_1 - \tilde{\phi}_2$ yields $\|\mathbf{DADw}\| \ll 1$

$$\|\mathbf{DAD}^{-1}\| = \max_{\mathbf{v} \neq 0} \frac{\|\mathbf{v}\|}{\|\mathbf{DADv}\|} \geq \frac{\|\mathbf{w}\|}{\|\mathbf{DADw}\|} \gg 1$$

Quasi linear dependencies on nonsmooth bases

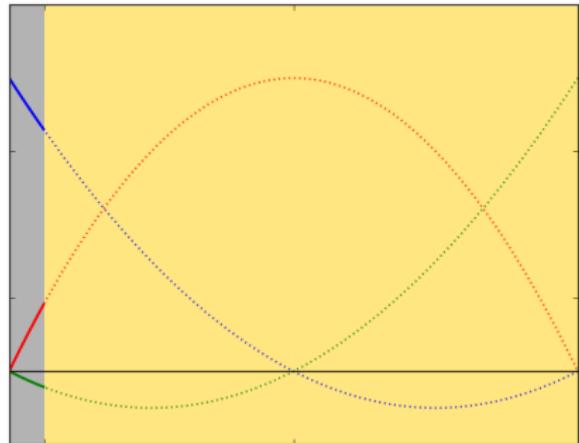


Original basis Φ

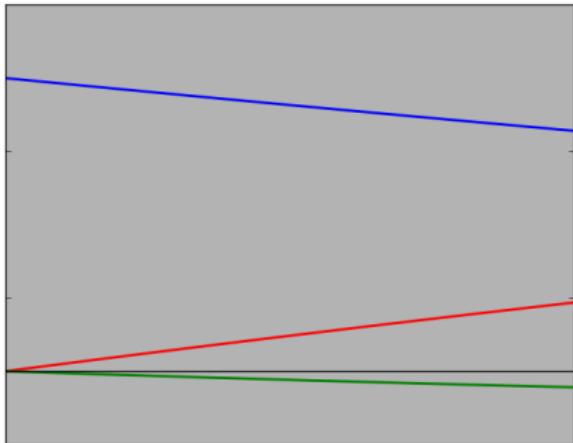
Quasi linear dependencies on nonsmooth bases

Quasi linear dependencies are a frequent phenomenon on high order bases with low regularity!

Quasi linear dependencies on nonsmooth bases



Original basis Φ

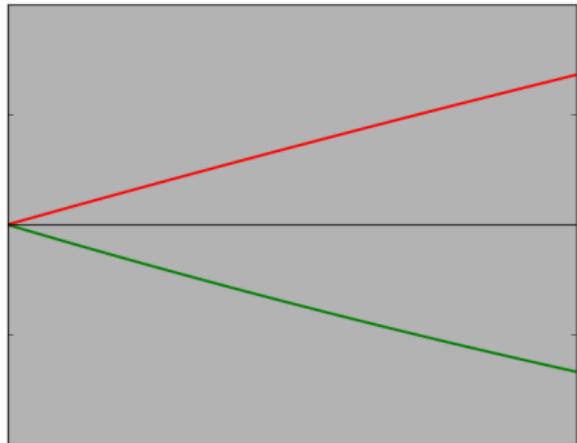


Restricted basis Φ

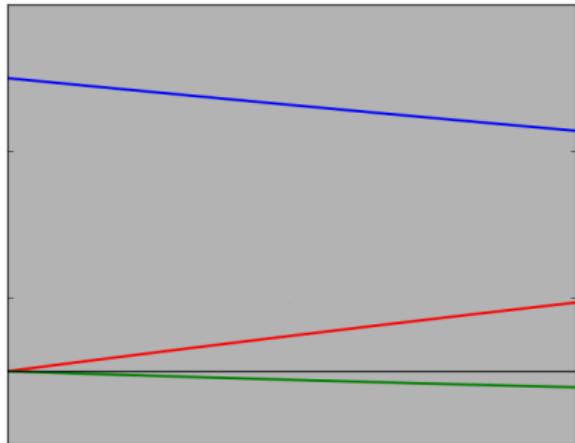
Quasi linear dependencies on nonsmooth bases

Quasi linear dependencies are a frequent phenomenon on high order bases with low regularity!

Quasi linear dependencies on nonsmooth bases



Scaled basis $\tilde{\Phi} = \mathbf{D}\Phi$

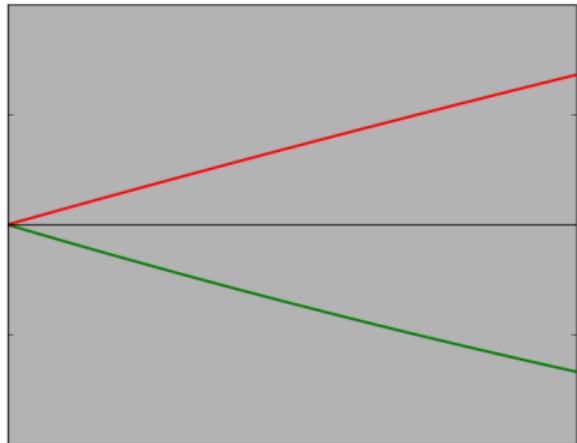


Restricted basis Φ

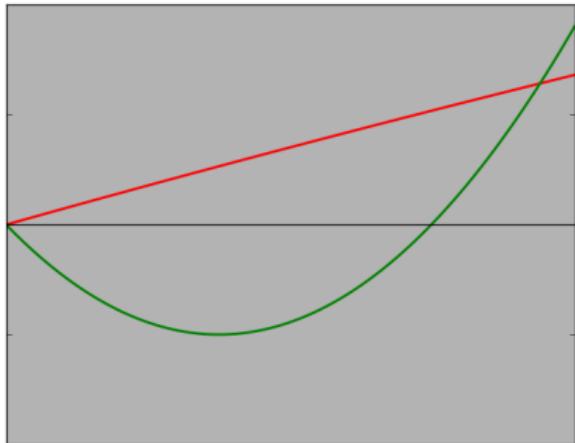
Quasi linear dependencies on nonsmooth bases

Quasi linear dependencies are a frequent phenomenon on high order bases with low regularity!

Quasi linear dependencies on nonsmooth bases



Scaled basis $\tilde{\Phi} = \mathbf{D}\Phi$



Orthonormalized basis $\bar{\Phi} = \mathbf{S}\Phi$

Quasi linear dependencies on nonsmooth bases

Quasi linear dependencies are a frequent phenomenon on high order bases with low regularity!

Generalization for non-SPD problems

$$\mathbf{S} = \begin{bmatrix} \textcolor{yellow}{\bullet} & & & & \\ \textcolor{yellow}{\bullet} & \textcolor{yellow}{\bullet} & & & \\ & & \textcolor{black}{\bullet} & & \\ & & & \textcolor{black}{\bullet} & \\ & & & & \textcolor{black}{\bullet} \\ & & & & & \vdots \\ & & & & & \textcolor{green}{\bullet} \\ & & & & & \textcolor{green}{\bullet} \\ & & & & & \textcolor{green}{\bullet} \end{bmatrix}$$

Interpretation

$\mathbf{S}^T \mathbf{S}$ is equal to the inverse of $\mathcal{R}(\mathbf{A})$, which is the restriction of matrix \mathbf{A} to its diagonal and the blocks of quasi linear dependent functions

Generalization for non-SPD problems

$$\mathbf{S}^T \mathbf{S} = \begin{bmatrix} \textcolor{blue}{\bullet} & \textcolor{blue}{\bullet} \\ \textcolor{blue}{\bullet} & \textcolor{blue}{\bullet} \\ & & \textcolor{black}{\bullet} \\ & & & \textcolor{black}{\bullet} \\ & & & & \textcolor{black}{\bullet} \\ & & & & & \vdots & \vdots \\ & & & & & \textcolor{red}{\bullet} & \textcolor{red}{\bullet} \end{bmatrix}$$

Interpretation

$\mathbf{S}^T \mathbf{S}$ is equal to the inverse of $\mathcal{R}(\mathbf{A})$, which is the restriction of matrix \mathbf{A} to its diagonal and the blocks of quasi linear dependent functions

Generalization for non-SPD problems

$$\mathbf{S}^T \mathbf{S} = \begin{bmatrix} \textcolor{yellow}{\bullet} & \textcolor{yellow}{\bullet} \\ \textcolor{yellow}{\bullet} & \textcolor{yellow}{\bullet} \\ & \cdot \\ & \vdots & \vdots \\ & \textcolor{green}{\bullet} & \textcolor{green}{\bullet} \\ & \textcolor{green}{\bullet} & \textcolor{green}{\bullet} \end{bmatrix} = \mathcal{R}(\mathbf{A})^{-1}$$

Interpretation

$\mathbf{S}^T \mathbf{S}$ is equal to the inverse of $\mathcal{R}(\mathbf{A})$, which is the restriction of matrix \mathbf{A} to its diagonal and the blocks of quasi linear dependent functions

Generalization for non-SPD problems

$$\mathbf{S}^T \mathbf{S} = \begin{bmatrix} \textcolor{blue}{\bullet} & \textcolor{blue}{\bullet} & \textcolor{blue}{\bullet} \\ \textcolor{blue}{\bullet} & \textcolor{blue}{\bullet} & \textcolor{blue}{\bullet} \\ \textcolor{blue}{\bullet} & \textcolor{blue}{\bullet} & \textcolor{blue}{\bullet} \\ & \ddots & & \textcolor{black}{\bullet} \\ & & \textcolor{black}{\bullet} & & \textcolor{black}{\bullet} \\ & & & \textcolor{red}{\bullet} & \textcolor{red}{\bullet} & \textcolor{red}{\bullet} \\ & & & \textcolor{red}{\bullet} & \textcolor{red}{\bullet} & \textcolor{red}{\bullet} \\ & & & \textcolor{red}{\bullet} & \textcolor{red}{\bullet} & \textcolor{red}{\bullet} \end{bmatrix} = \mathcal{R}(\mathbf{A})^{-1}$$

Interpretation

$\mathbf{S}^T \mathbf{S}$ is equal to the inverse of $\mathcal{R}(\mathbf{A})$, which is the restriction of matrix \mathbf{A} to its diagonal and the blocks of quasi linear dependent functions

Generalization for non-SPD problems

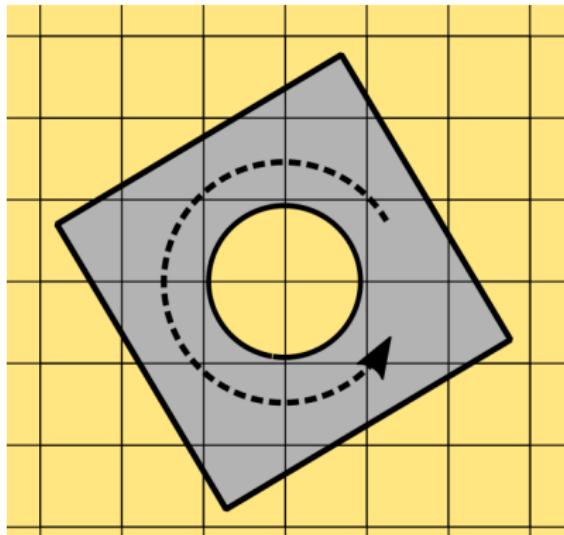
$$\mathbf{S}^T \mathbf{S} = \begin{bmatrix} \textcolor{yellow}{\bullet} & \textcolor{yellow}{\bullet} & \textcolor{yellow}{\bullet} \\ \textcolor{yellow}{\bullet} & \textcolor{yellow}{\bullet} & \textcolor{yellow}{\bullet} \\ \textcolor{yellow}{\bullet} & \textcolor{yellow}{\bullet} & \textcolor{orange}{\bullet} & \textcolor{red}{\bullet} & \textcolor{red}{\bullet} \\ & \textcolor{red}{\bullet} & \textcolor{red}{\bullet} & \textcolor{red}{\bullet} \\ & \textcolor{red}{\bullet} & \textcolor{red}{\bullet} & \textcolor{red}{\bullet} \\ & & & \bullet \\ & & & \textcolor{green}{\bullet} & \textcolor{green}{\bullet} & \textcolor{green}{\bullet} \\ & & & \textcolor{green}{\bullet} & \textcolor{green}{\bullet} & \textcolor{green}{\bullet} \\ & & & \textcolor{green}{\bullet} & \textcolor{green}{\bullet} & \textcolor{green}{\bullet} \end{bmatrix} \neq \mathcal{R}(\mathbf{A})^{-1}$$

Interpretation

Additive-Schwarz preconditioning

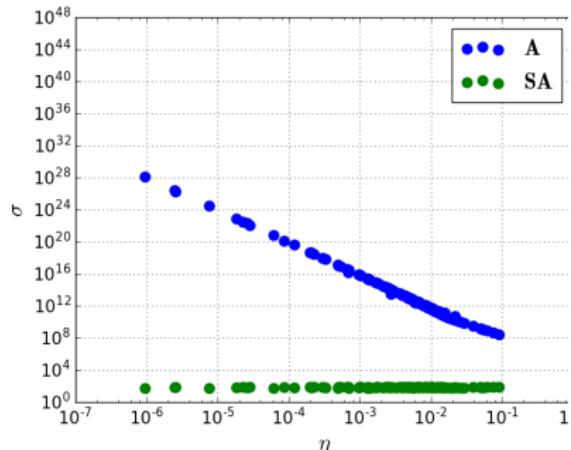
Results for flow problems

Domain:

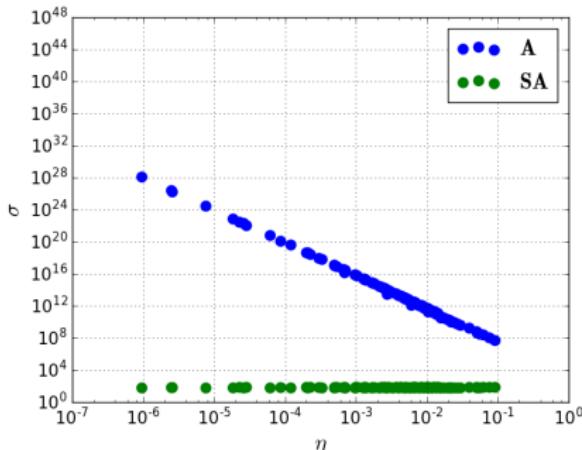


Results for flow problems

$$p = 2$$



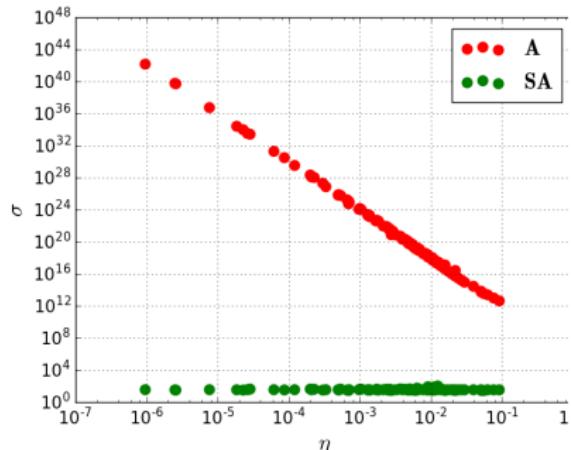
Stokes



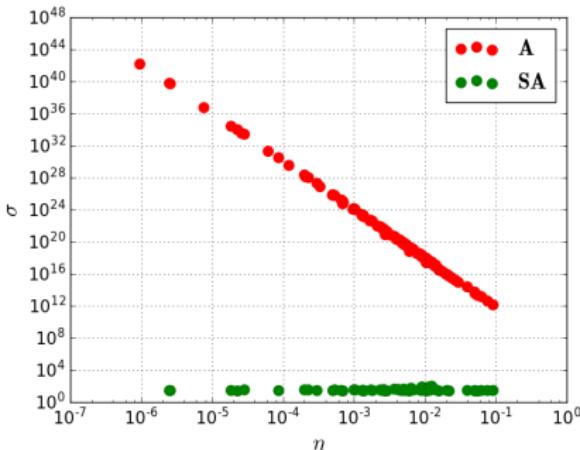
Navier-Stokes

Results for flow problems

$$p = 3$$

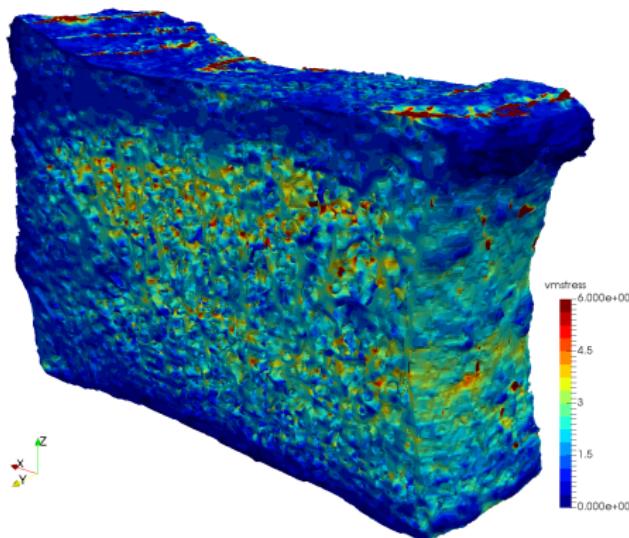


Stokes



Navier-Stokes

Results for elasticity problems



CT-scan of human
vertebra

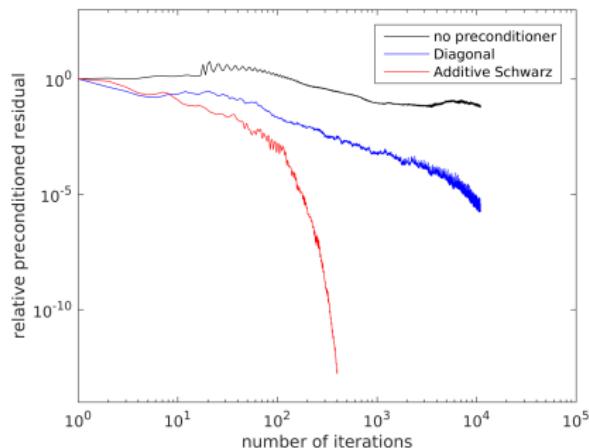
John Jomo
Collaboration with: Stefan Kollmannsberger
Ernst Rank



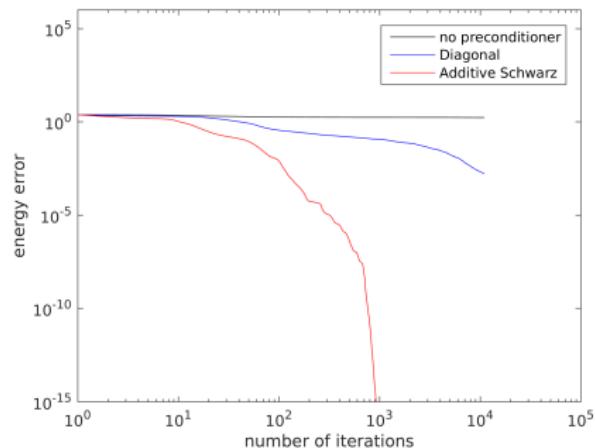
Technische Universität München

Results for elasticity problems

relative preconditioned residual



energy error



John Jomo

Collaboration with: Stefan Kollmannsberger

Ernst Rank



Technische Universität München

Conclusion

Summary

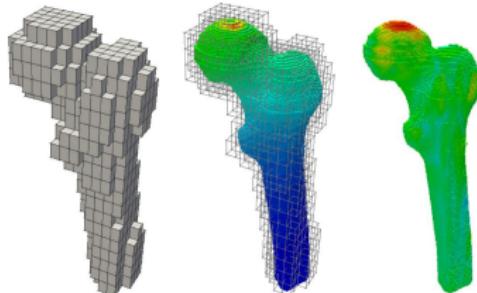
- Introduction to immersed finite elements methods
- Conditioning analysis
- Effective tailored preconditioner

Future work (in immersed methods)

- Preconditioning
 - Combinations with other (multigrid) preconditioners
 - Parallel and meshless implementations
- Explicit dynamics
- Compatible (divergence free) discretizations
- Multiphase flows

Advanced School on

Immersed Methods



06-09 | 2017
November | 2017



Topics

- Fundamental modelling aspects
- Boundary and coupling conditions
- Numerical integration techniques
- Ghost penalty
- Conditioning and solution methods
- Image-based modelling
- Application in fluid and solid mechanics
- Application in isogeometric analysis
- Application in topology optimization

Registration before October 31st 2017
Website: www.tue.nl/emiworkshop
Contact: emi@tue.nl

Eindhoven University of
Technology



Lecturers

- Alexander Düster
Hamburg University of Technology
Mats Larson
Umeå University
Ernst Rank
Technical University of Munich
Martin Ruess
University of Glasgow
Ole Sigmund
Technical University of Denmark
Clemens Verhoosel
Eindhoven University of Technology