TUG

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# Physics based block preconditioning with sparse approximate inverses in MueLu: An application to beam/solid interaction



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2. Problem description and discretization

### 3. Multigrid for fiber/solid systems

Classification of MG for coupled problems

Block preconditioning for fiber/solid systems

### 4. Sparse Approximate Inverses

Approximation Sparsity pattern selection SPAI smoother

### 5. Numerical experiments

Weak scaling study



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- Beam / Solid interactions occur in a wide variety of scenarios:
  - Engineering (steel-reinforced concrete, composite materials)
  - Biomechanics (collagen fibers in connective tissue)
- Time-to-solution dominated by cost for linear solver
  - Scalability through multilevel methods
  - Algebraic Multigrid (AMG) for its flexibility
  - But: Ill-conditioned matrix due to discretization and penalty regularization prohibit out-of-the-box block smoothing



### Goal

Scalable AMG preconditioner for beam / solid interaction problems in penalty formulation

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$$\begin{pmatrix} \mathbf{K}_{B} + \epsilon \mathbf{D}^{T} \kappa^{-1} \mathbf{D} & -\epsilon \mathbf{D}^{T} \kappa^{-1} \mathbf{M} \\ -\epsilon \mathbf{M}^{T} \kappa^{-1} \mathbf{D} & \mathbf{K}_{S} + \epsilon \mathbf{M}^{T} \kappa^{-1} \mathbf{M} \end{pmatrix} \begin{pmatrix} \Delta \mathbf{d}_{B} \\ \Delta \mathbf{d}_{S} \end{pmatrix} = - \begin{pmatrix} \mathbf{r}_{B} \\ \mathbf{r}_{S} \end{pmatrix}$$

### Legend

- (.)<sub>s</sub> solid contribution
- (.)<sub>B</sub> beam contribution
- d displacement DOFs
- **r** residual
- $\epsilon$  penalty parameter
- $\kappa$  scaling factor



Fiber/solid coupling in penalty formulation results in a linear system with  $2 \times 2$  block structure:

$$\begin{pmatrix} \mathbf{K}_{B} + \epsilon \mathbf{D}^{T} \kappa^{-1} \mathbf{D} & -\epsilon \mathbf{D}^{T} \kappa^{-1} \mathbf{M} \\ -\epsilon \mathbf{M}^{T} \kappa^{-1} \mathbf{D} & \mathbf{K}_{S} + \epsilon \mathbf{M}^{T} \kappa^{-1} \mathbf{M} \end{pmatrix} \begin{pmatrix} \Delta \mathbf{d}_{B} \\ \Delta \mathbf{d}_{S} \end{pmatrix} = - \begin{pmatrix} \mathbf{r}_{B} \\ \mathbf{r}_{S} \end{pmatrix}$$

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- Beam DOFs
- Solid DOFs
- Coupling constraints



# Coupled system of equations

Fiber/solid coupling in penalty formulation results in a linear system with  $2 \times 2$  block structure:

$$\begin{pmatrix} \mathbf{K}_{B} + \epsilon \mathbf{D}^{T} \kappa^{-1} \mathbf{D} & -\epsilon \mathbf{D}^{T} \kappa^{-1} \mathbf{M} \\ -\epsilon \mathbf{M}^{T} \kappa^{-1} \mathbf{D} & \mathbf{K}_{S} + \epsilon \mathbf{M}^{T} \kappa^{-1} \mathbf{M} \end{pmatrix} \begin{pmatrix} \Delta \mathbf{d}_{B} \\ \Delta \mathbf{d}_{S} \end{pmatrix} = - \begin{pmatrix} \mathbf{r}_{B} \\ \mathbf{r}_{S} \end{pmatrix}$$



### **Challenges:**

- Highly non-diagonal dominant and ill-conditioned block matrix due to penalty regularization
- Block matrix may be nonsymmetric due to beam formulation

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### **Block-iterative MG:**

# If MG possible for all matrix blocks:



- Coupling only on fine level
- Independent MG hierarchies to approximate each block inverse

### Fully coupled MG:



- Include coupling into all MG levels
- Requires near nullspace for all blocks

# Multigrid for block matrices



### **Block-iterative MG:**

If MG possible for all matrix blocks:



- Coupling only on fine level
- Independent MG hierarchies to approximate each block inverse

If MG impossible for one matrix block:



- Coupling only on fine level
- MG to approximate one block inverse
- Approximate the other block inverse w/o MG

### Fully coupled MG:



- Include coupling into all MG levels
- Requires near nullspace for all blocks



### Matrices arising from solid discretizations:

- Near nullspace easy to compute
- AMG readily available and well established in literature

### Matrices arising from beam discretizations:

- Construction of near nullspace depends on beam formulation (work in progress)
- Coarsening of short fibers with just a few nodes not very sensible
- AMG (or MG) for beams not available in literature so far

### Our approach: Block-iterative scheme with aggregation-based AMG for solid block



- Smoothed-Aggregation AMG (SA-AMG)<sup>1</sup> for solid block
  - One-level method to approximate inverse of beam block
- ▶ Penalty destroys diagonal dominance ~→ no block relaxation methods
- Requires block methods to address 2 × 2 block structure

### <sup>1</sup>Vanek1996a

### **Starting point:**

- ▶ 2 × 2 blocking of system matrix
- Block relaxation (e.g. Block Gauss-Seidel) does not work (lack of diagonal dominance)
- System matrix potentially nonsymmetric

### **Remedy:**

- Rely on Schur complements
  - Block LU
  - Uzawa
  - SIMPLE
- Use AMG for selected block inverses







# Example: SIMPLE smoother<sup>2</sup>



### Task

Solve 
$$\begin{pmatrix} A & B_1^T \\ B_2 & C \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix}$$

### Ideas

$$\begin{pmatrix} A & B_1^T \\ B_2 & C \end{pmatrix} \approx \begin{pmatrix} A & A \widehat{A}^{-1} B_1^T \\ B_2 & C \end{pmatrix} =$$

Use splitting

$$\begin{pmatrix} A & \\ B_2 & C - B_2 \widehat{A}^{-1} B_1^T \end{pmatrix} \begin{pmatrix} I & \frac{1}{\omega} \widehat{A}^{-1} B_1^T \\ & \frac{1}{\omega} I \end{pmatrix} \begin{pmatrix} I & \\ & \omega I \end{pmatrix}$$

with  $\widehat{A}$  being an easy-to-invert approximation of A.

- CheapSIMPLE: Use cheap smoothing methods for displacement prediction and SchurComplement equation
- Many known variants: SIMPLEC, SIMPLER...

### Algorithm

### 1. Calculate residual

$$\begin{pmatrix} r_1 \\ r_2 \end{pmatrix} = \begin{pmatrix} b_1 \\ b_2 \end{pmatrix} - \begin{pmatrix} A & B_1^T \\ B_2 & C \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}^b$$

2. Solve for prediction of beam field:

$$A\Delta \widetilde{x_1} = r_1$$

3. Solve "SchurComplement" equation:

$$\left(C - B_2 \widehat{A}^{-1} B_1^T\right) \Delta \widetilde{x_2} = r_2 - B_2 \Delta \widetilde{x_1}$$

4. Update step:

$$\Delta \widehat{x_2} = \omega \Delta \widetilde{x_2}$$
$$\Delta \widehat{x_1} = \Delta \widetilde{x_1} - \frac{1}{\omega} \widehat{A}^{-1} B_1^T \Delta \widehat{x_2}$$

5. Increment  $k \rightarrow k + 1$ :

$$\begin{pmatrix} x_1 \\ x_2 \end{pmatrix}^{k+1} = \begin{pmatrix} x_1 \\ x_2 \end{pmatrix}^k + \begin{pmatrix} \Delta \widehat{x_1} \\ \Delta \widehat{x_2} \end{pmatrix}$$

### <sup>1</sup>Patankar1972a

# Approximations in Schur complement preconditioners

### Challenge

### How to approximate the Schur complement?

- 1. Approximation  $\widehat{A} \approx A^{-1}$  to form Schur complement S
  - $\Rightarrow$  Governed by the type of block method
  - $\Rightarrow$  e.g.  $\widehat{A} := diag(A)^{-1}$
- 2. Approximate block inverses within Schur complement preconditioner by standard AMG
  - $\Rightarrow~$  Approximation quality can be controlled through the AMG settings

### Complication

Simple (e.g. diagonal) approximations of the inverse inside the Schur complement calculation is insufficient, because the penalty regularization acts also matrix elements far away from the diagonal.

### Remedy

Construct and use a sparse approximate inverse to form the Schur complement.



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- Use matrix graph J(A) to calculate apprimated inverse A on this sparsity pattern
- Key idea: minimization of Frobenius norm:

 $\min_{\widehat{A}\in\Sigma}||A\widehat{A}-I||_F$ 

with  $\Sigma$  being the set of all sparse matrices with some known structure

### **Parallel computation**

Inherent parallelism through decomposition into row-wise independent least squares problems:  $||\widehat{AA} - I||_F^2 = \sum_{k=1}^n ||(\widehat{AA_k} - I)e_k||_2^2$ , for each row k solve  $\min_{\widehat{A_k}} ||\widehat{AA_k} - e_k||_2$ with QR-decomposition

### <sup>2</sup>Grote1997a



### Some observations

- Using just the pattern of A as input might not result in a satisfactory result.
- The matrix pattern needs to be enriched for a good sparse inverse approximation.
- Combining rows of graph J(A) such that<sup>4</sup>:  $J(A_k^l) = J(A_k^{l-1})J(A^{l-1})$
- Pre- and post filtering of input graph and sparse inverse approximation with threshold value τ

### SPAI with static pattern selection

- 1. Tresholding of *J*(*A*)
- 2. Determine graph of powers of A:  $J(A^l)$
- 3. Calculate sparse inverse approximation  $\widehat{A}$
- **4**. Post filtering of  $\widehat{A}$

### 4Chow2001a



### Note

Not all block inverses need to be approximated with an AMG V-cycle.

### Prediction of beam solution:

- Due to Schur complement calculation, good approximation of inverse of beam matrix block is already available
- Using this information for smoothing results in the following SPAI smoother<sup>5</sup>:

 $x^{k+1} = x^k - \widehat{A}(Ax^k - b)$ 

with  $\widehat{A}$  being a sparse approximate inverse

### Schur complement equation:

Solve with conventional AMG method:

- Standard relaxation methods don't converge due to non-diagonal dominance
- Polynomial smoothers like the Chebychev iteration provide decent results, but are not very robust.
- ILU as a smoother works well, thought setup cost is high.

### <sup>5</sup>Broeker2002a



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### Settings

### Discretization

# Solid DOFs: 27783 # Beam DOFs: 1548 # procs: 1

### Solver

Newton tolerance: 10<sup>-6</sup> (rel) BiCGSTAB tolerance: 10<sup>-8</sup> (rel) MG preconditioner: 3 level SA-AMG, ILU(1), LU

### Material parameters

Solid: $E_S = 1 \frac{N}{m^2}, \nu_S = 0.3$ <br/>hyperelastic Saint Venant-Kirchhoff modelBeam: $E_B = 10 \frac{N}{m^2}, \nu_B = 0.0$ <br/>torsion-free Kirchhoff-Love modelPenalty: $\epsilon = 10 \frac{N}{m}$ 



- minimal working problem to be used for weak scaling study
- bottom surface is fixed, tensile surface load on top surface

# A first attempt on weak scaling I

### Weak scaling study: Cube filled with randomly placed and oriented fibers.







### Weak scaling hierachy

ID	n <sup>proc</sup>	n <sup>S</sup> <sub>DOF</sub>	$n_{DOF}^{B}$	n <sup>total</sup> DOF	n <sup>total</sup> DOF/proc
1	1	27783	1548	29331	29331.0
2	8	206763	14544	221307	27663.4
3	27	680943	52188	733131	27153.0
4	64	1594323	124788	1719111	26861.1
5	125	3090903	247560	3338463	26707.7
6	216	5314683	432300	5746983	26606.4
7	343	8409663	688956	9098619	26526.6
8	512	12519843	1035876	13555719	26476.1
9	729	17789223	1484736	19273959	26438.9
10	1000	24361803	2037192	26398995	26399.0



Domaindecompositionapproach based on a geometricbisection for ID = 2 with  $n^{proc} = 8$ 

# A first attempt on weak scaling III





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# Summary



### **Applications:**

- Mixed-dimensional fiber/solid coupling with penalty constraint enforcement
- For now, only torsion-free Kirchhoff–Love beam elements
  - Sufficient for a broad range of applications
  - Restriction to straight center line in reference configuration

# Multigrid block preconditioner:

- Block preconditioning based on Schur complement
  - Sparse approximate inverse for approximation of Schur complement
  - SA-AMG to invert Schur complement
- Successful proof of concept for engineering applications
- First weak scaling results



# Outlook



# Formulation

- Extend to other beam formulations, especially Simo-Reissner
- Include rotational coupling between beams and solid
- Extend to other coupling scenarios, e.g. beam/solid contact
- Performance
  - Thread parallelism for SPAI computation
  - Reduce preconditioner setup time
  - Investigate and improve weak scaling behavior

### AMG(BlockMethod)



- Consider coupling constraints on all levels
- Assembly of the beam DOFs nullspace specific to beam formulation
- Re-use SPAI as level smoother

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# Thank you!

### **Collaborators:**

- Matthias Mayr, UniBw M
- Ivo Steinbrecher, UniBw M
- Alexander Popp, UniBw M

### **References:**

Open-source implementation will be available in Trilinos/MueLu: https://trilinos.github.io/muelu.html

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dtec.bw: Digitalization and Technology Research Center of the Bundeswehr through the project hpc.bw: Competence Platform for High Performance Computing

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