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# Teko Usage in Aria (UUR)

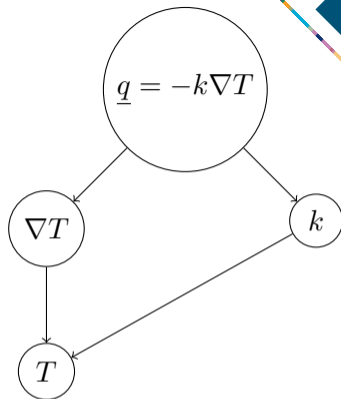
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SAND2024-14144C



## SIERRA/Aria:

- Expression-based multi-physics simulation code<sup>1</sup>
- Segregated and monolithic physics coupling
- Monolithic solver/preconditioner approaches:
  - KLU2/SuperLU sparse direct solvers through Amesos2
  - Domain-decomposition with incomplete LU (DD-ILU) through Ifpack2
  - *Teko physics-based preconditioners*
- Teko solvers integrate with existing Trilinos packages:
  - Amesos2, Belos, Ifpack2, MueLu



<sup>1</sup>Notz, Pawlowski, and Sutherland, “Graph-based software design for managing complexity and enabling concurrency in multiphysics PDE software”.

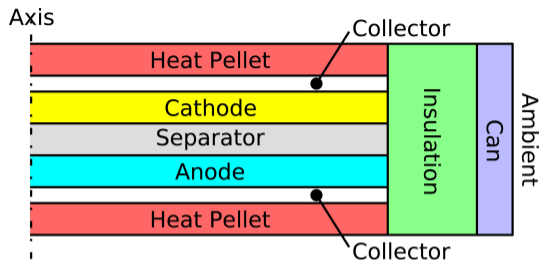


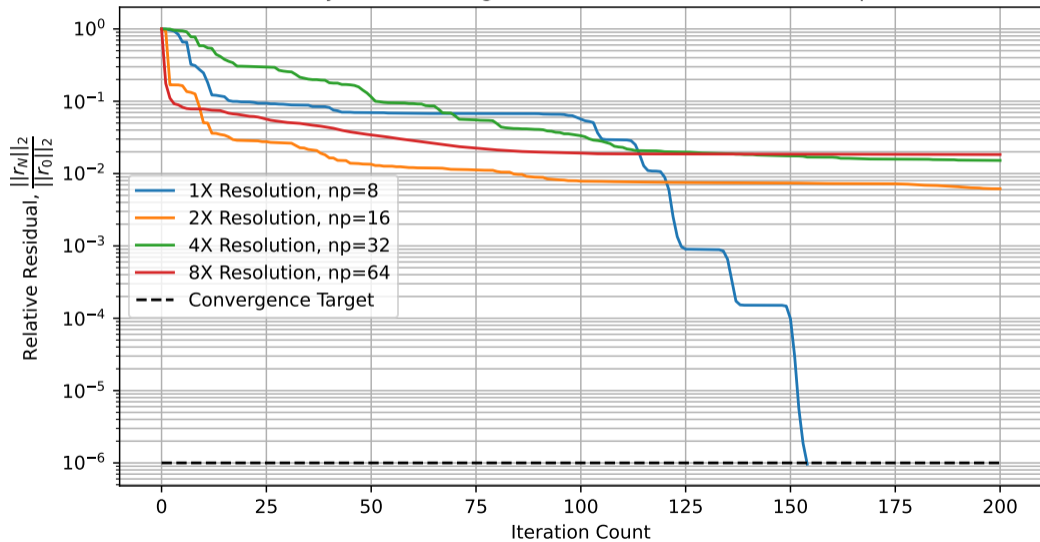
Figure: 2D axisymmetric multi-physics simulation domain<sup>1</sup>

- Includes several multi-physics couplings<sup>1</sup>:
  - Butler-Volmer
  - Stefan-Maxwell
  - Darcy's Law
  - Continuity
- DD-ILU may not converge
- Sparse direct solvers do not scale
- *Weak Scaling*: 57,640 DOFs to 461,120 DOFs
  - Solver: DD-ILU(1) preconditioned GMRES(200)

<sup>1</sup>Voskuilen, Moffat, Schroeder, and Roberts, "Multi-fidelity electrochemical modeling of thermally activated battery cells".



### Battery Weak Scaling, GMRES(200), DD-ILU(1), Overlap=1





- Teko block splitting (trial-and-error):
  - $c$ : solid phase species
  - $x$ : liquid phase mass fractions
  - $p$ : liquid phase pressure
  - $v$ : solid/liquid phase voltage
- Block representation of matrix:

$$\mathcal{A} = \begin{bmatrix} A_{cc} & A_{cx} & A_{cp} & A_{cv} \\ A_{xc} & A_{xx} & A_{xp} & A_{xv} \\ A_{pc} & A_{px} & A_{pp} & A_{pv} \\ A_{vc} & A_{vx} & A_{vp} & A_{vv} \end{bmatrix}$$

- Construct Block Gauss-Seidel preconditioner:

$$\tilde{\mathcal{M}}^{-1}(\mathcal{A}) = \begin{bmatrix} A_{cc} & A_{cx} & A_{cp} & A_{cv} \\ & A_{xx} & A_{xp} & A_{xv} \\ & & A_{pp} & A_{pv} \\ & & & A_{vv} \end{bmatrix}^{-1}$$



- $\tilde{\mathcal{M}}^{-1}(\mathcal{A})$  requires sub-block inverses, replace with  $\mathcal{M}^{-1}(\mathcal{A})$ :

$$\mathcal{M}^{-1}(\mathcal{A}) = \begin{bmatrix} M_{cc} & A_{cx} & A_{cp} & A_{cv} \\ & M_{xx} & A_{xp} & A_{xv} \\ & & M_{pp} & A_{pv} \\ & & & M_{vv} \end{bmatrix}^{-1}$$

## Convergence Result

A single resistant sub-block solver can derail the entire solver:

$$\begin{aligned} \kappa(\mathcal{M}^{-1}(\mathcal{A})\mathcal{A}) &\leq \kappa(\mathcal{M}^{-1}(\mathcal{A})\tilde{\mathcal{M}}(\mathcal{A})) \cdot \kappa(\tilde{\mathcal{M}}^{-1}(\mathcal{A})\mathcal{A}) \\ &\geq \underbrace{\max_{\forall M_{ii}^{-1}A_{ii}} (\kappa(M_{ii}^{-1}A_{ii}))}_{\text{Sub-block Solver Conditioning}} \cdot \underbrace{\kappa(\tilde{\mathcal{M}}^{-1}(\mathcal{A})\mathcal{A})}_{\text{Multi-physics Coupling}} \end{aligned}$$



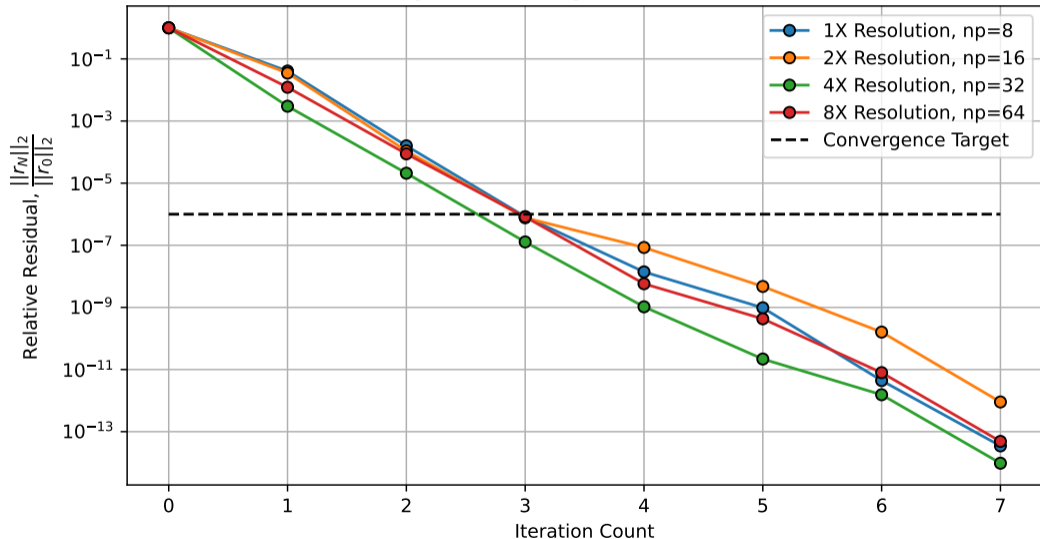
- *A single resistant sub-block solver can derail the entire solver*
- Use iterative sub-block solves to ensure convergence
  - Preconditioner  $M^{-1}(\mathcal{A})$  *changes* per iteration
  - Requires *flexible* GMRES (F-GMRES)<sup>2</sup>
  - *Same orthogonalization cost*
  - *Double restart memory*
- Approximate sub-block inverses for diagonal entries:
  - $M_{cc}^{-1}$ : Jacobi
  - $M_{xx}^{-1}$ : GMRES + DD-ILU
  - $M_{pp}^{-1}$ : GMRES + DD-ILU
  - $M_{vv}^{-1} = A_{vv}^{-1}$ : KLU2 sparse direct solver

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<sup>2</sup>Saad, “A flexible inner-outer preconditioned GMRES algorithm”.



### Battery Weak Scaling, F-GMRES(200), Teko







- Teko provides effective preconditioners for multi-physics problems
  - Especially useful when DD-ILU struggles

## Ingredients for Teko Solver Setup

Teko solver setup requires:

1. Physics-to-sub-block mapping
2. Ordering sub-blocks
3. Solvers/preconditioners for each sub-block

- *Goal:* provide ability to *auto-magically* generate reasonable Teko settings

```
1  begin tpetra equation solver teko_linear_solver
2      begin preset solver
3          solver type = teko_multiphysics
4      end preset solver
5  end tpetra equation solver
```



## Ordering sub-blocks:

- Block Gauss-Seidel ordering matters:

$$\kappa \left( \underbrace{\begin{bmatrix} A & B \\ & D \end{bmatrix}^{-1}}_{\tilde{\mathcal{M}}^{-1}(\mathcal{A})} \underbrace{\begin{bmatrix} A & B \\ C & D \end{bmatrix}}_A \right) \neq \kappa \left( \underbrace{\begin{bmatrix} D & C \\ & A \end{bmatrix}^{-1}}_{\tilde{\mathcal{M}}^{-1}(\mathcal{RAC})} \underbrace{\begin{bmatrix} D & C \\ B & A \end{bmatrix}}_{\mathcal{RAC}} \right)$$

## Optimization Problem

Find block ordering permutation  $\mathcal{R}^* (\cdot) \mathcal{C}^*$  such that:

$$(\mathcal{R}^*, \mathcal{C}^*) = \arg \min_{\forall \mathcal{R}, \mathcal{C}} \kappa(\tilde{\mathcal{M}}^{-1}(\mathcal{RAC}) \mathcal{RAC})$$

- Naïve brute-force approach to optimal ordering is exponential in  $n_b$
- *Collaboration with SandiaAI*: use graph-based heuristic for ordering



# Ordering Heuristic

**Input** :  $\mathcal{A}$  with  $n_b$  blocks,  $n$  maximum iterations

**Output**: Re-ordered  $\mathcal{A}'$  more suitable for block Gauss-Seidel

```

1  $\mathcal{A}^{(1)} = \mathcal{A}$ 
2 for  $k \leftarrow 1$  to  $n$  do
3   Construct undirected graph  $\mathcal{G}^{(k)} \leftarrow (\mathcal{V} = \{1, \dots, n_b\}, \mathcal{E} = (i, j) \mid \forall i \in \mathcal{V}, j \geq i)$ 
4   Construct symmetric edge-weight matrix  $W_{i,j}^{(k)} \leftarrow \begin{cases} \left\| \mathcal{A}_{i,j}^{(k)} \right\|_F & i > j \\ \left\| \mathcal{A}_{j,i}^{(k)} \right\|_F & j > i \\ 0 & i = j \end{cases}$ 
5    $D^{(k)} \leftarrow \text{diag}(D_1^{(k)}, \dots, D_{n_b}^{(k)})$  with  $D_i^{(k)} = \sum_{j=1}^{n_b} W_{i,j}^{(k)}$ 
6   Form symmetric weighted graph Laplacian  $L^{(k)} \leftarrow D^{(k)} - W^{(k)}$ 
7    $Q^{(k)} \Lambda^{(k)} (Q^{(k)})^T \leftarrow L^{(k)}$  with  $\lambda_1 \leq \lambda_2 \leq \dots \leq \lambda_{n_b}$ 
8   Construct re-ordering from Fiedler vector  $\mathcal{R}^{(k)} (\cdot) \mathcal{C}^{(k)} \leftarrow \text{argsort}(V[:,2])$  // Second smallest
9    $\mathcal{A}^{(k+1)} \leftarrow \mathcal{R}^{(k)} \mathcal{A}^{(k)} \mathcal{C}^{(k)}$ 
10  if  $\mathcal{A}^{(k+1)} = \mathcal{A}^{(k)}$  then return  $\mathcal{A}^{(k+1)}$ 
11 end
12 return  $\mathcal{A}^{(n+1)}$ 

```



- Apply ordering heuristic to battery problem
  - $c$ : solid phase species
  - $x$ : liquid phase mass fractions
  - $p$ : liquid phase pressure
  - $v$ : solid/liquid phase voltage
- Previous ordering:

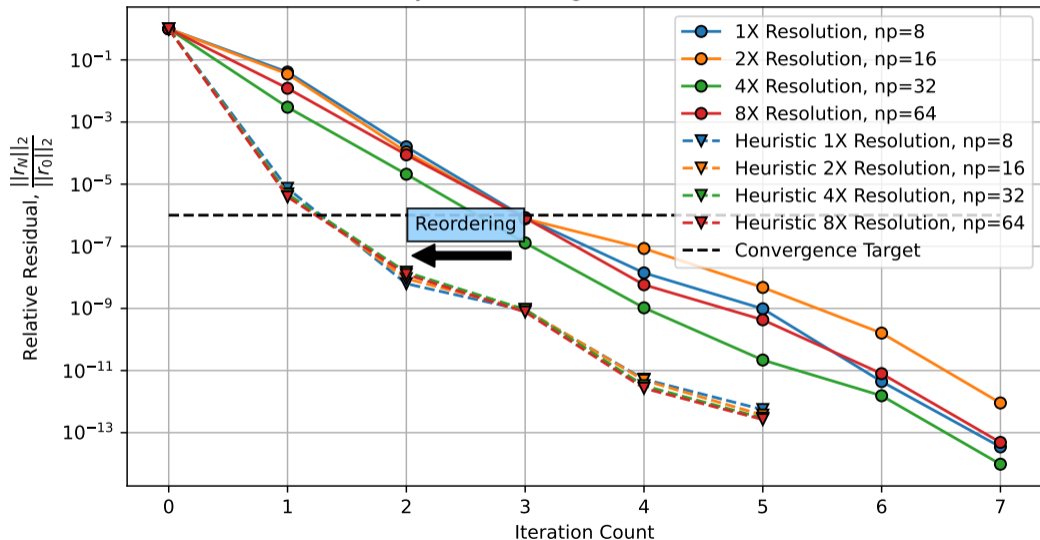
$$\mathcal{A} = \begin{bmatrix} A_{cc} & A_{cx} & A_{cp} & A_{cv} \\ A_{xc} & A_{xx} & A_{xp} & A_{xv} \\ A_{pc} & A_{px} & A_{pp} & A_{pv} \\ A_{vc} & A_{vx} & A_{vp} & A_{vv} \end{bmatrix}, \mathcal{M}^{-1}(\mathcal{A}) = \begin{bmatrix} M_{cc} & A_{cx} & A_{cp} & A_{cv} \\ & M_{xx} & A_{xp} & A_{xv} \\ & & M_{pp} & A_{pv} \\ & & & M_{vv} \end{bmatrix}^{-1}$$

- New ordering:

$$\mathcal{A}' = \begin{bmatrix} A_{vv} & A_{vp} & A_{vc} & A_{vx} \\ A_{pv} & A_{pp} & A_{pc} & A_{px} \\ A_{cv} & A_{cp} & A_{cc} & A_{cx} \\ A_{xv} & A_{xp} & A_{xc} & A_{xx} \end{bmatrix}, \mathcal{M}^{-1}(\mathcal{A}') = \begin{bmatrix} M_{vv} & A_{vp} & A_{vc} & A_{vx} \\ & M_{pp} & A_{pc} & A_{px} \\ & & M_{cc} & A_{cx} \\ & & & M_{xx} \end{bmatrix}^{-1}$$



### Battery Weak Scaling, F-GMRES(200), Teko





Physics-to-sub-block mapping:

- Strongly coupled physics may require monolithic approach:

$$\kappa \left( \underbrace{\begin{bmatrix} A & B & C \\ & E & F \\ & & J \end{bmatrix}^{-1}}_{\tilde{M}^{-1}(A)} \underbrace{\begin{bmatrix} A & B & C \\ D & E & F \\ G & H & J \end{bmatrix}}_A \right) \neq \kappa \left( \underbrace{\begin{bmatrix} \begin{bmatrix} A & B \\ D & E \end{bmatrix} & \begin{bmatrix} C \\ F \\ J \end{bmatrix} \end{bmatrix}^{-1}}_{\tilde{M}^{-1}(A')} \underbrace{\begin{bmatrix} \begin{bmatrix} A & B \\ D & E \\ G & H \end{bmatrix} & \begin{bmatrix} C \\ F \\ J \end{bmatrix} \end{bmatrix}}_{A'} \right)$$

- Monolithically treat combined block with DD-ILU
  - Alternative: sub-iterate via *Hierarchical Block Gauss-Seidel*
- Simple greedy heuristic for grouping (next slide)
- At every step, reduce  $n_b \times n_b$  block system to  $(n_b - 1) \times (n_b - 1)$ :

$$\mathcal{A}_{(n_b-1) \times (n_b-1)}^* \leftarrow \arg \min_{\mathcal{A}'_{(n_b-1) \times (n_b-1)}} \max_{\forall i, j \in \{1, \dots, n_b-1\}, j \neq i} \left\| (\mathcal{A}'_{i,j})_{(n_b-1) \times (n_b-1)} \right\|_F$$

## Block Grouping Heuristic

**Input** : Matrix  $\mathcal{A}$  with  $n_b$  blocks, maximum iterations  $n < n_b$ , threshold  $\tau$

**Output**: Re-grouped matrix  $\mathcal{A}'$  with  $n'_b < n_b$  blocks

/\* Note:  $\left\| \begin{bmatrix} A & B \\ C & D \end{bmatrix} \right\|_F = \sqrt{\|A\|_F^2 + \|B\|_F^2 + \|C\|_F^2 + \|D\|_F^2}$ . \*/

```
1  $\mathcal{A}^{(0)} \leftarrow \mathcal{A}$ 
2  $r_0 \leftarrow \max_{i,j,i \neq j} \|\mathcal{A}_{i,j}^{(0)}\|_F$ 
3 for  $k \leftarrow 0$  to  $n - 1$  do
4    $(i^*, j^*) \leftarrow \arg \max_{i,j,i \neq j} \|\mathcal{A}_{i,j}^{(k)}\|_F$ 
5    $\mathcal{A}^{(k+1)} \leftarrow \text{combine}(\mathcal{A}^{(k)}, i^*, j^*)$  //  $\mathcal{A}^{(k+1)}$  is a  $n_b - (k + 1)$  block system
6    $r_{k+1} \leftarrow \max_{i,j,i \neq j} \|\mathcal{A}_{i,j}^{(k+1)}\|_F$ 
7   if  $\frac{r_{k+1}}{r_0} \leq \tau$  then return  $\mathcal{A}^{(k+1)}$ 
8 end
9 return  $\mathcal{A}^{(n)}$ 
```



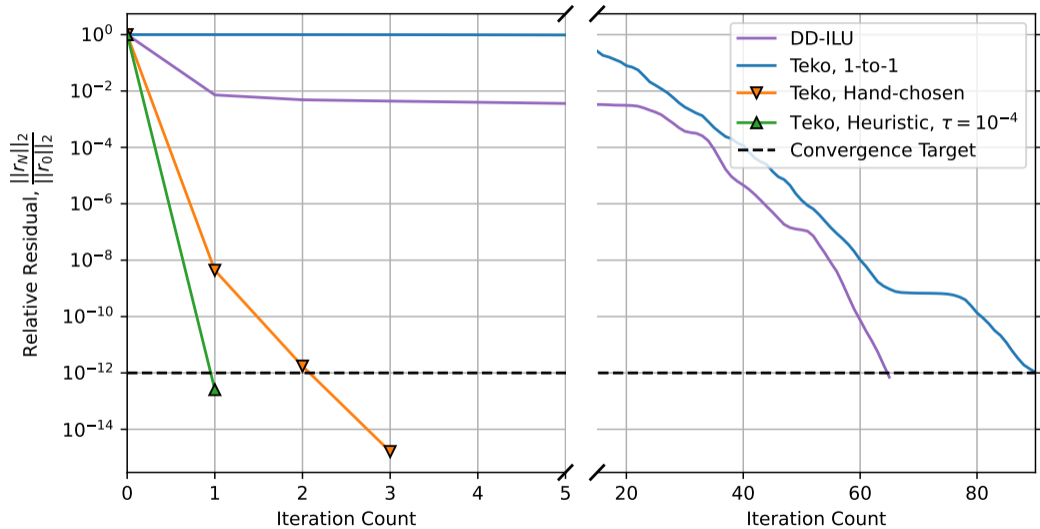
OMD FIC multi-physics example:

- 14 coupled PDEs:
  - Solid conduction
  - Porous-fluid coupled flow
  - Enthalpy
  - Species transport
- Comparison between four preconditioners with  $\frac{\|r_N\|_2}{\|r_0\|_2} = 10^{-12}$ 
  - Monolithic DD-ILU
  - One-to-one physics-to-block mapping, using heuristic order
  - Hand-chosen grouping/order
  - Heuristic grouping/ordering, target reduction  $\tau = 10^{-4}$
- KLU2 sparse direct solver for sub-blocks





# OMD FIC Linear Solver, GMRES(200)





Solvers/preconditioners for each sub-block:

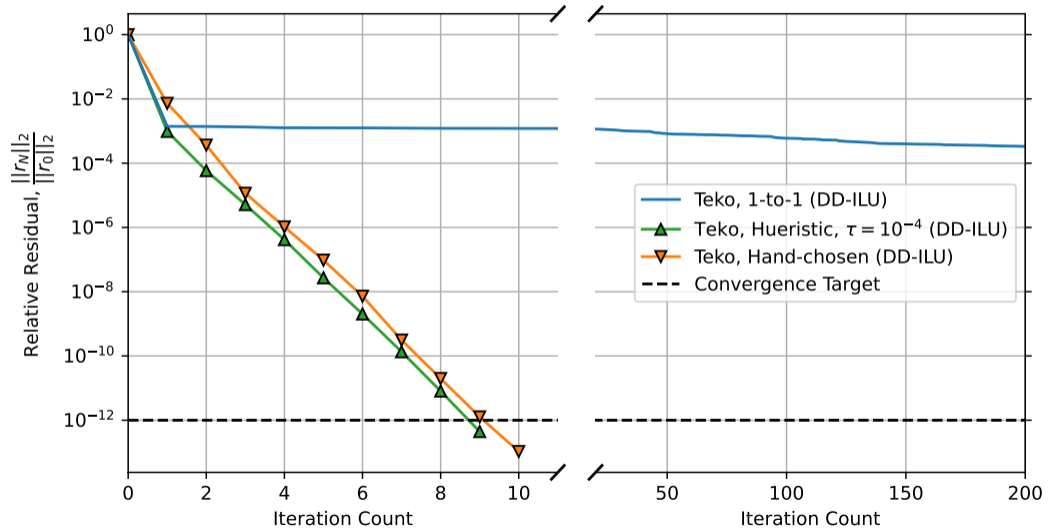
- Many sub-blocks will incorporate several physics

Poor man's solution

- DD-ILU preconditioned GMRES(30)
- Target one order-of-magnitude residual reduction
- F-GMRES(200) as outer solver
  
- Repeat OMD FIC example:
  - One-to-one physics-to-block mapping, using heuristic order
  - Hand-chosen grouping/order
  - Heuristic grouping/ordering, target reduction  $\tau = 10^{-4}$



### OMD FIC Linear Solver, F-GMRES(200)



### Concluding remarks:

- DD-ILU is a great work-horse preconditioner
  - *It can fail to converge*
- Teko provides diversification to multi-physics solver portfolio
- Relies on existing preconditioners/solvers at sub-block level:
  - DD-ILU
  - MueLu
- Flexible, extensible package with many options
- Large landscape of solver settings difficult for users to navigate
  - *WIP: Provide users with guidance through heuristics*

### Questions?

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