

# Recent Algorithmic (and Practical) Developments in ML

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# Outline

- Introduction to ML.
- Solving Maxwell's Equations w/ RefMaxwell.
- Repartitioning w/ Zoltan and Hypergraphs.
- Conclusions & Future Work.



# Trilinos Summary

## Core

Teuchos

Zoltan

Thyra

RTOp

Epetra

EpetraExt

ForTrilinos

Isorropia

## Discretizations

Intrepid

phdMesh

Rythmos

## Solvers

AztecOO

ML

NOX

Meros

LOCA

Ifpack

Amesos

Anasazi

Belos

Moocho

## Methods

Moertel

Sacado



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# ML Features (1)

- ML provides scalable multilevel/multigrid preconditioners.
- Method types
  - Smoothed Aggregation (SA) - symmetric or nearly symmetric problems.
  - Non-symmetric SA - non-symmetric problems.
  - MatrixFree - matrix-free SA.
  - DD / DD-ML - domain decomposition.
  - Maxwell - Maxwell's equations.
  - RefMaxwell - new method for Maxwell's equations.



# ML Features (2)

- Simple Trilinos interface.
- Teuchos::ParameterList driven options.  
Has sensible defaults (override what you don't like).
- Parameter validation for accuracy.
- MATLAB interface for some features (MLMEX).



# Using ML

```
// Start with a problem & build solver
Epetra_LinearProblem Problem(A, &LHS, &RHS);
AztecOO solver(Problem);

// Override any defaults
Teuchos::ParameterList List;
List.set("smoother: sweeps",2);

// Build the preconditioner
MultiLevelPreconditioner Prec(A,List);
solver.SetPrecOperator(Prec);

solver.Iterate(100,1e-12); // Solve
```

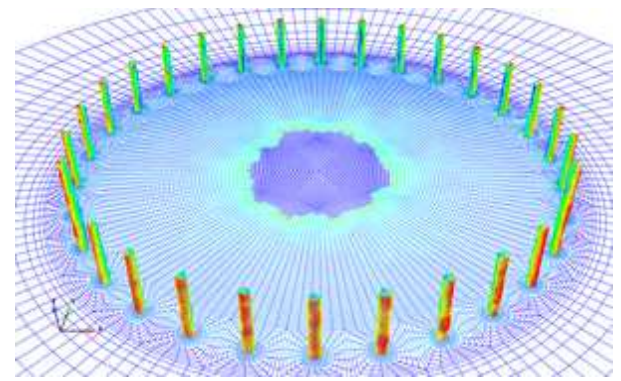


# Outline

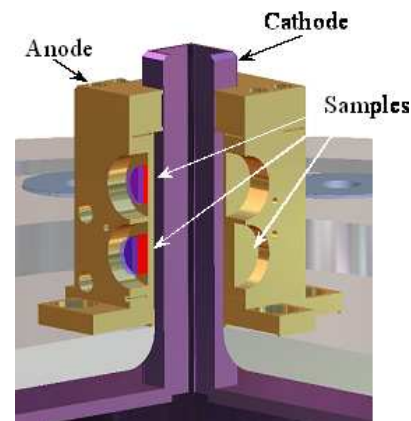
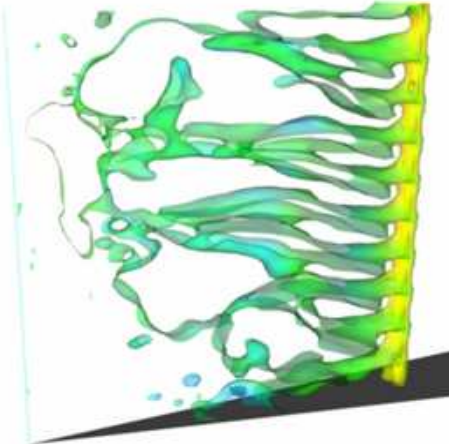
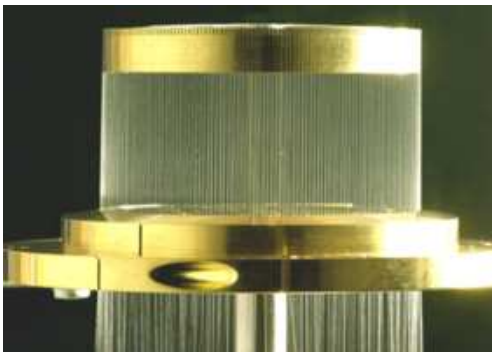
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# Target Applications

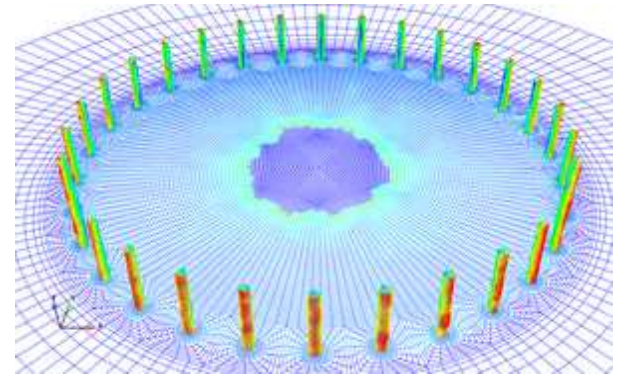


- Electromagnetic phenomena modeled by Maxwell's equations occur in many Sandia applications.
- HEDP: Wire arrays and liners for Z machine simulations.
- Magnetic Launch: Coil & rail guns (ONR).
- Code: ALEGRA & Trilinos/ML (SNL).

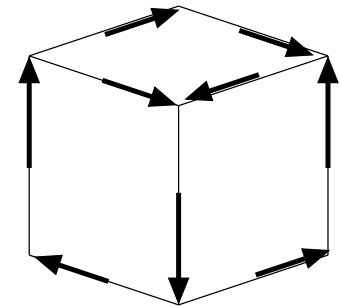


# Maxwell's Equations

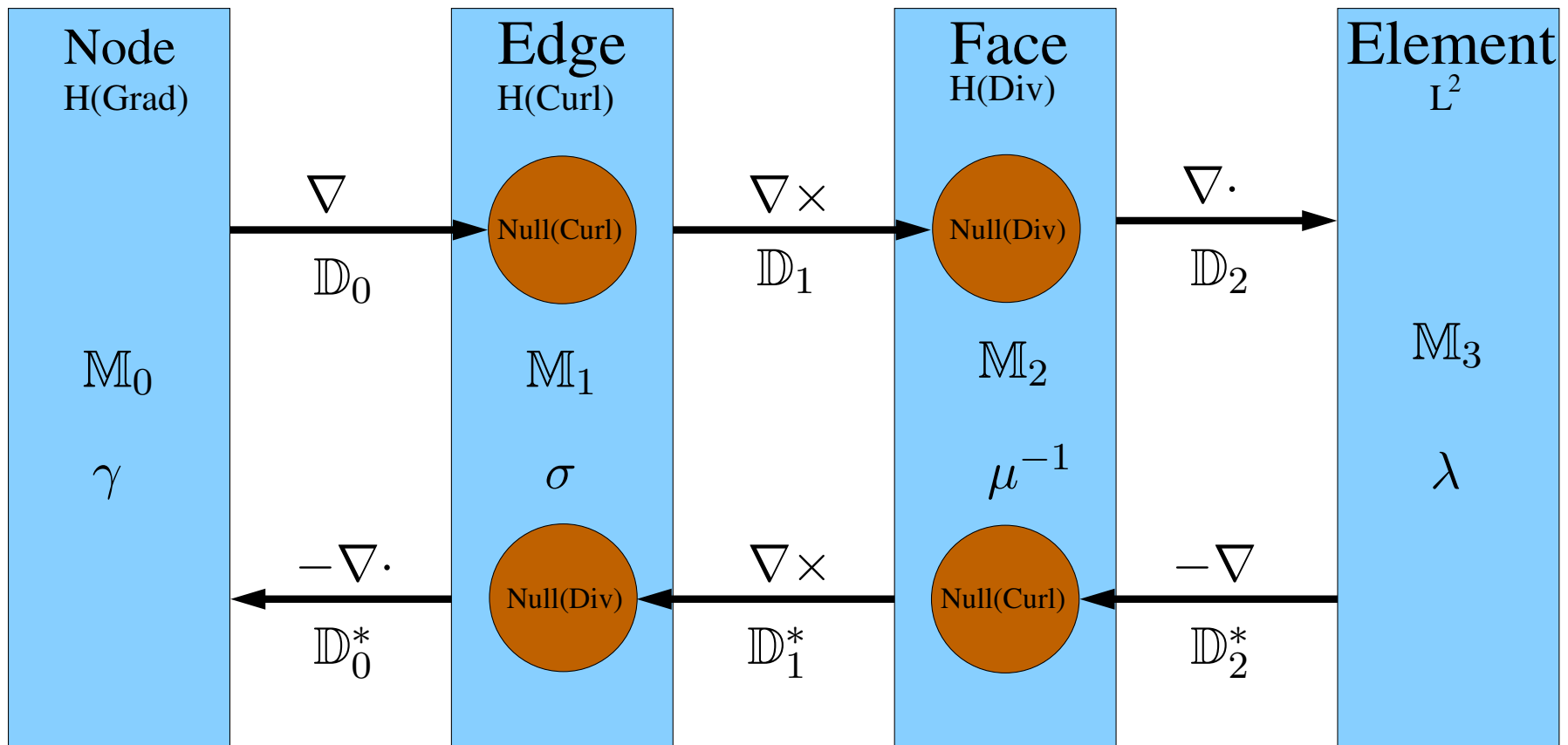
$$\begin{aligned}\nabla \times \frac{1}{\mu} \nabla \times \mathbf{E} + \sigma \mathbf{E} &= 0 \quad \text{in } \Omega \\ \mathbf{n} \times \mathbf{E} &= 0 \quad \text{on } \Gamma \\ \mathbf{n} \times \frac{1}{\mu} \nabla \times \mathbf{E} &= 0 \quad \text{on } \Gamma^*\end{aligned}$$



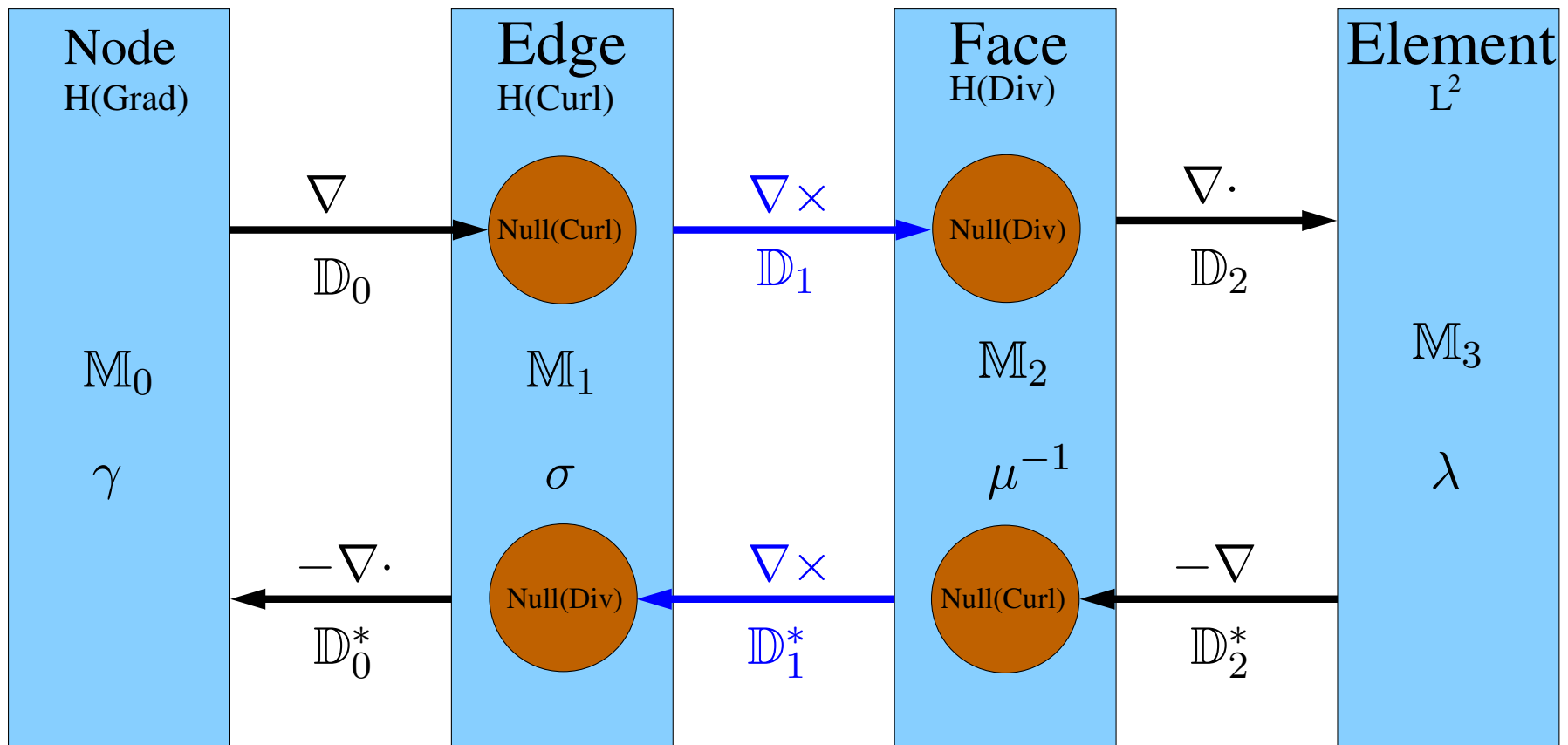
- $\nabla \times \nabla \phi = 0 \Rightarrow$  large null space complicates discretization + solver.
- Large jumps in  $\sigma$ .
- Significant mesh stretching.
- Large problems & repeated solves  
→ Scalable linear solvers are critical.



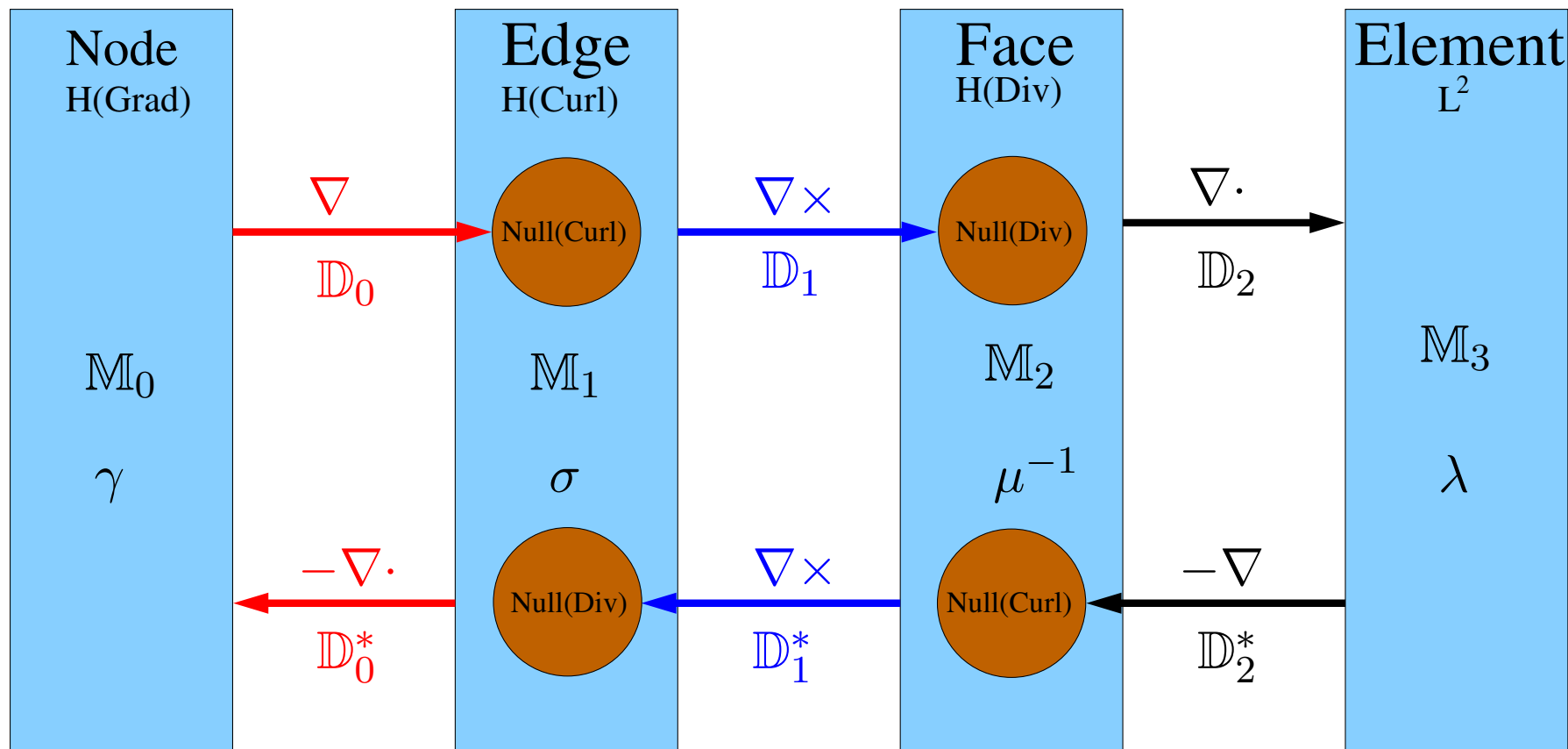
# Continuous/Discrete Relationship



# Continuous/Discrete Relationship



# Hodge Laplacian



•  $\Delta = \nabla \times \nabla \times + \nabla \nabla \cdot$

•  $L_1 = \mathbb{D}_1^* \mathbb{D}_1 + \mathbb{D}_0 \mathbb{D}_0^*$



# Discrete Hodge Decomposition

$$(\mathbb{M}_1 \mathbb{D}_1^* \mathbb{D}_1 + \mathbb{M}_1) e = b$$

- Consider

$$e = a + \mathbb{D}_0 p,$$

where  $\mathbb{D}_0^* a = 0$ .

- This gives us the block  $2 \times 2$  system

$$\begin{bmatrix} \mathbb{M}_1 \mathbb{D}_1^* \mathbb{D}_1 + \mathbb{M}_1 & \mathbb{M}_1 \mathbb{D}_0 \\ \mathbb{M}_0 \mathbb{D}_0^* & \mathbb{M}_0 \mathbb{D}_0^* \mathbb{D}_0 \end{bmatrix} \begin{bmatrix} a \\ p \end{bmatrix} = \begin{bmatrix} b \\ \mathbb{D}_0^T b \end{bmatrix}.$$



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$$\begin{bmatrix} \mathbb{M}_1 \mathbb{D}_1^* \mathbb{D}_1 + \mathbb{M}_1 + \mathbb{M}_1 \mathbb{D}_0 \mathbb{D}_0^* & \mathbb{M}_1 \mathbb{D}_0 \\ \mathbb{M}_0 \mathbb{D}_0^* & \mathbb{M}_0 \mathbb{D}_0^* \mathbb{D}_0 \end{bmatrix} \begin{bmatrix} a \\ p \end{bmatrix} = \begin{bmatrix} b \\ \mathbb{D}_0^T b \end{bmatrix}.$$

- We can add  $\mathbb{D}_0 \mathbb{D}_0^*$  w/o changing answer!



# Preconditioning

$$\begin{bmatrix} \mathbf{M}_1 \mathbf{D}_1^* \mathbf{D}_1 + \mathbf{M}_1 \mathbf{D}_0 \mathbf{D}_0^* + \mathbf{M}_1 & \mathbf{M}_1 \mathbf{D}_0 \\ \mathbf{M}_0 \mathbf{D}_0^* & \mathbf{M}_0 \mathbf{D}_0^* \mathbf{D}_0 \end{bmatrix} \begin{bmatrix} a \\ p \end{bmatrix} = \begin{bmatrix} b \\ \mathbb{D}_0^T b \end{bmatrix}.$$

- Use preconditioner:

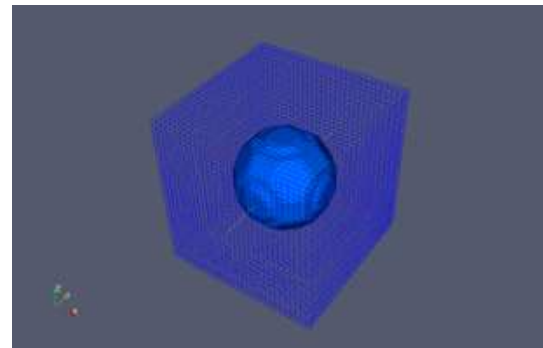
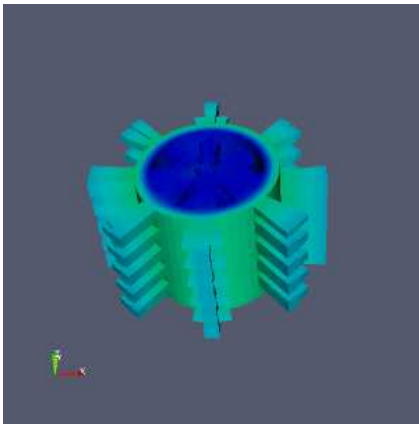
$$P^{-1} = \begin{bmatrix} I & \mathbb{D}_0 \end{bmatrix} \begin{bmatrix} \text{AMG}_{11} & \\ & \text{AMG}_{22} \end{bmatrix} \begin{bmatrix} I \\ \mathbb{D}_0^T \end{bmatrix}$$

- This preconditioner is implemented in ml/src/RefMaxwell in Trilinos 8.0.



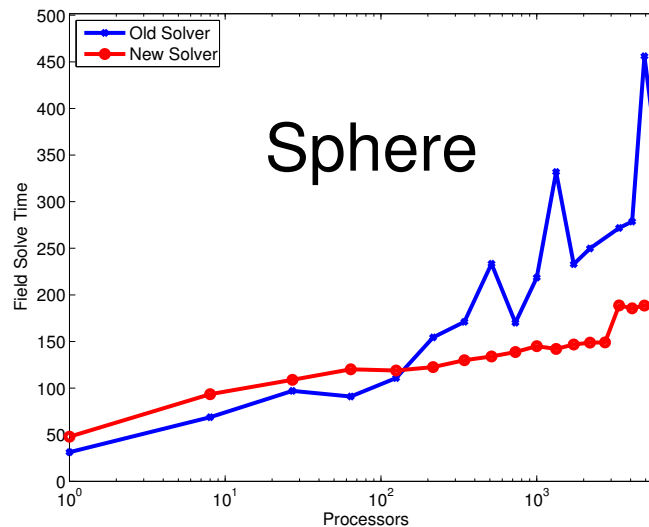
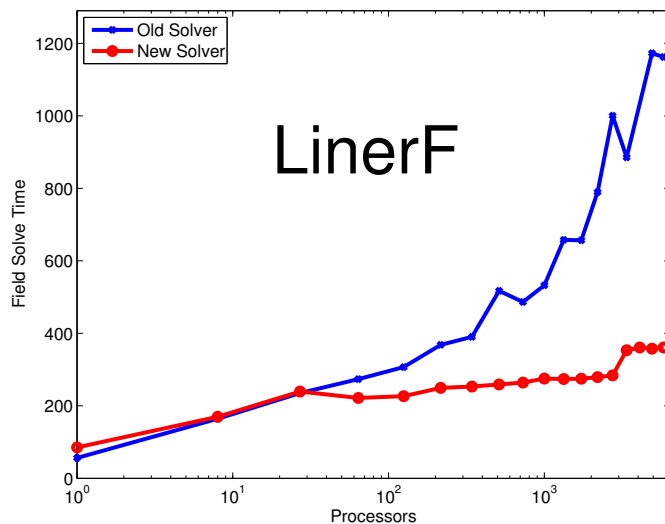
# 3D Weak Scaling

- Problem Code: ALEGRA (SNL).
- Problems: LinerF, Sphere.
- Material Parameters:  $1e6$  jump in conductivity.
- Geometry: Regular meshes.
- Compare **Maxwell** vs. **RefMaxwell**

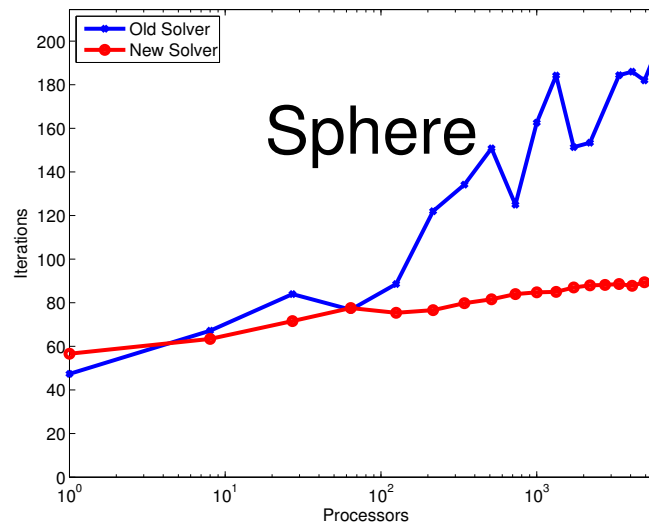
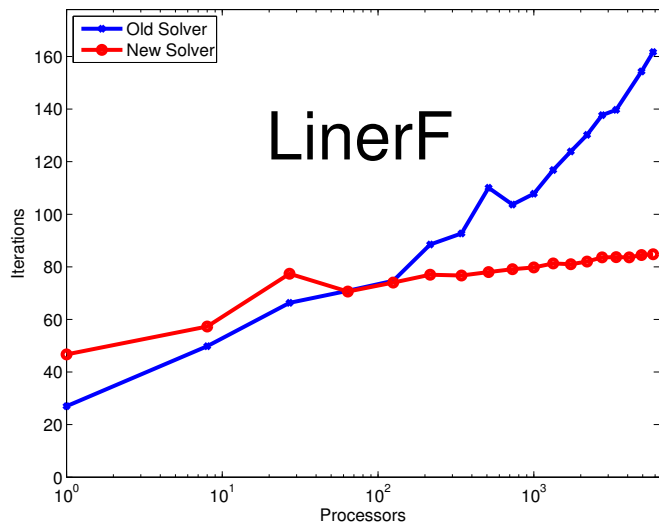


# Scaling: Old vs. New

Solve Time



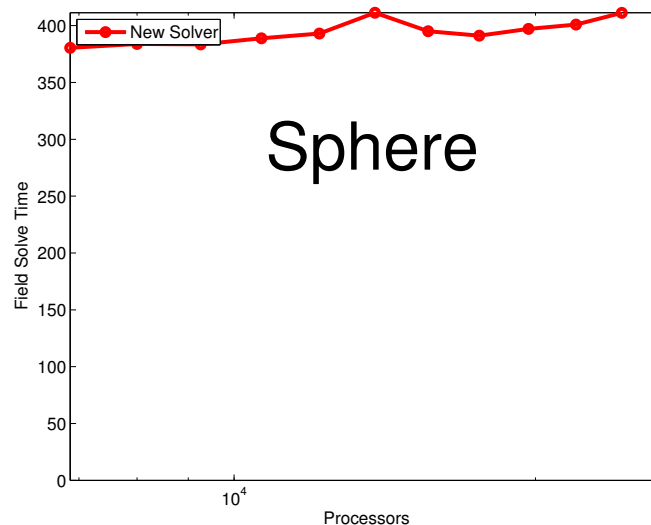
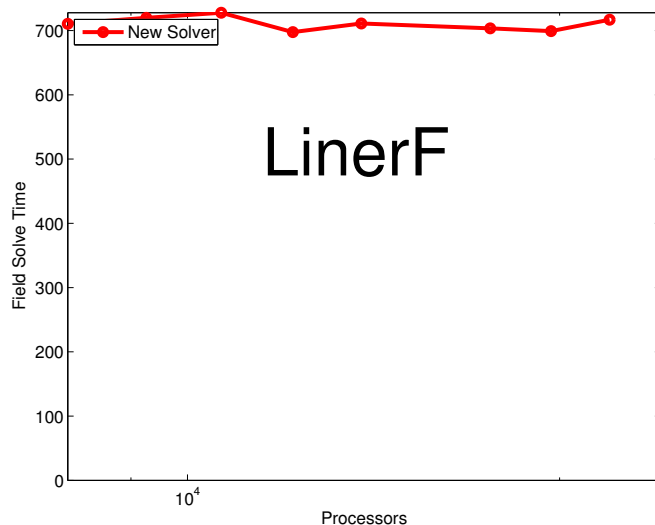
Iterations



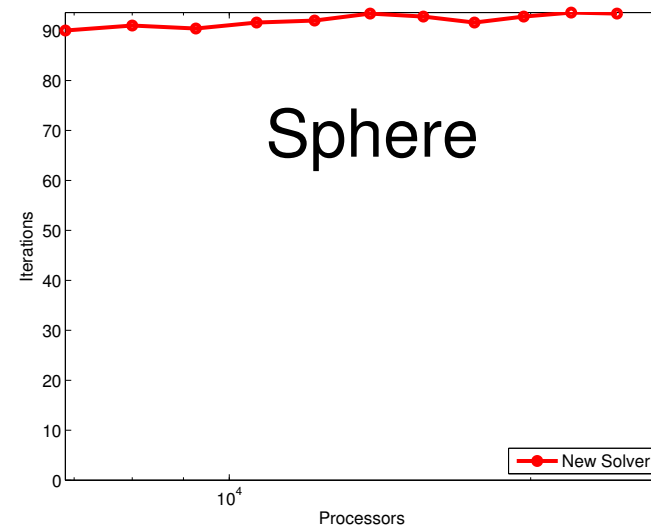
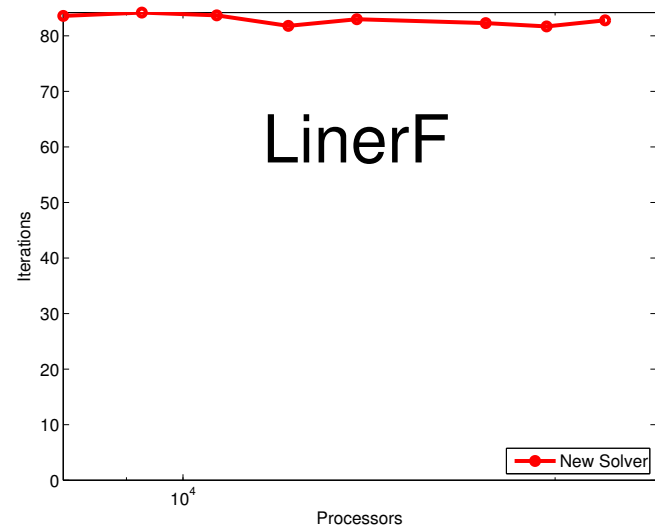
Number of Processors

# Jumbo Scaling: **New**

Solve Time



Iterations



Number of Processors



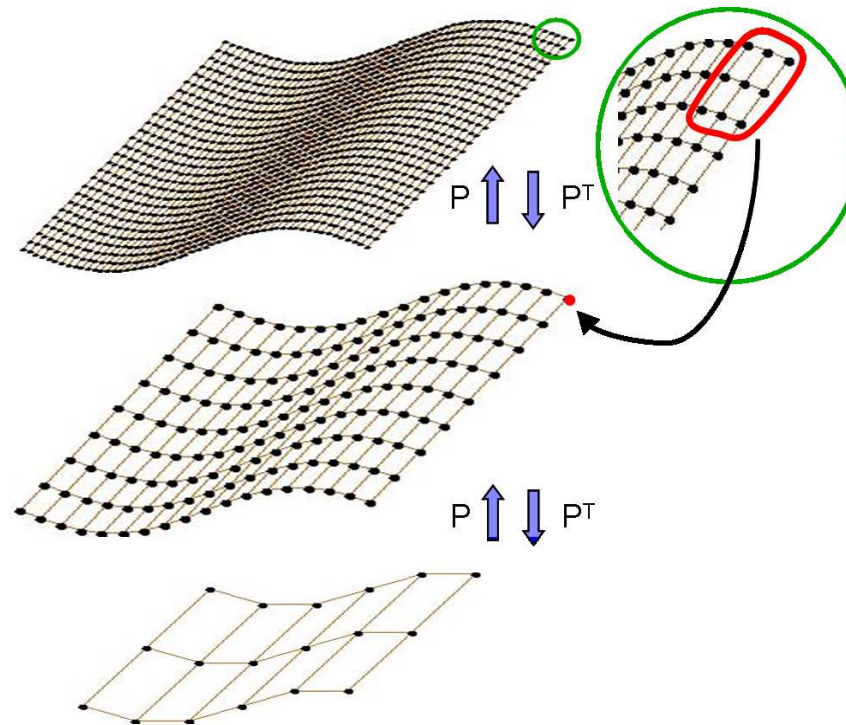
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# Why Repartitioning?

- Coarse grids  $\Rightarrow$  less work per proc  $\Rightarrow$  poor performance.
- Solution: Move data to leave some procs idle.

Computation  
Dominated



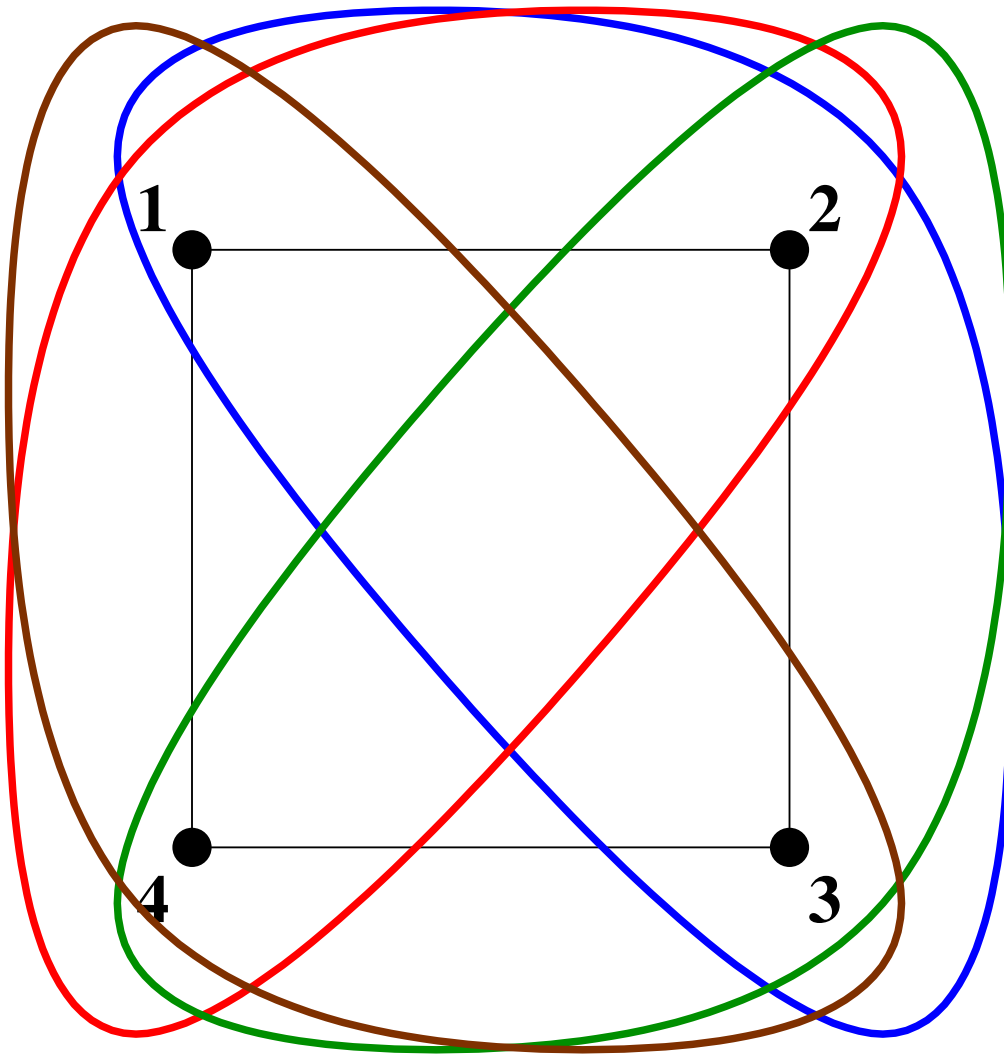
Communication  
Dominated



# Why Repartitioning?

- Goals
  - Move to a subset of processors  
⇒ Increase computation to communication ratio.
  - Rebalance load.
- Method
  - ML current supports RCB via Zoltan.
  - What about irregular meshes?
  - What about consistency between partitions at different levels?
  - New Feature: Hypergraph Repartitioning.  
⇒ To be released in Trilinos 9.0.

# What is a Hypergraph?



X	X		X
X	X	X	
	X	X	X
X		X	X

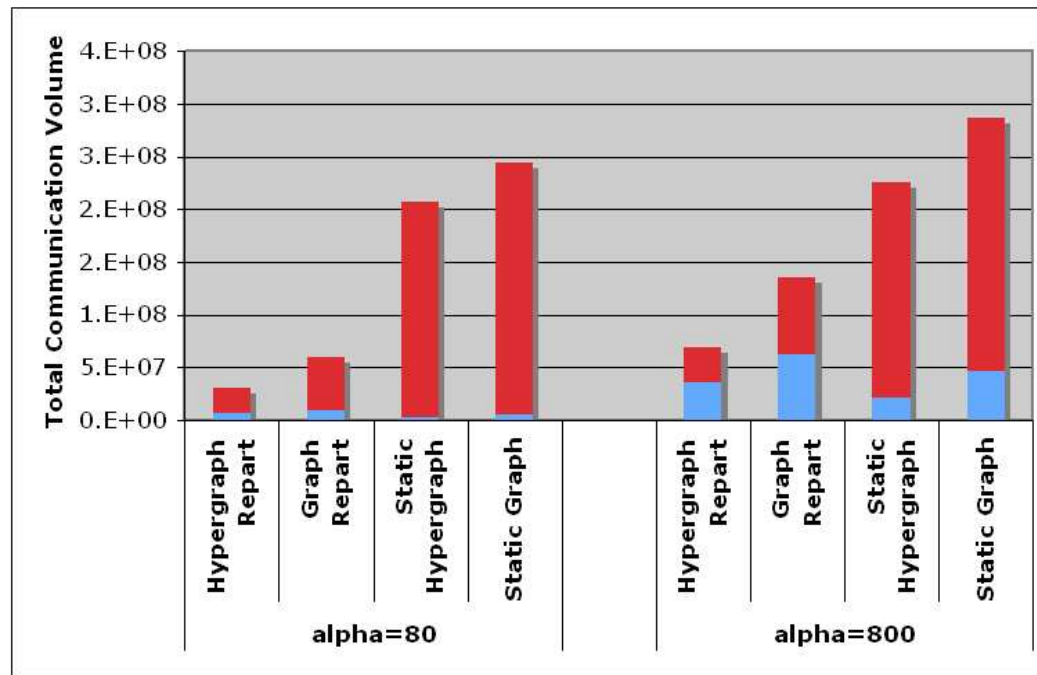


# Why Hypergraphs?

- For (compressed row) matrix:
  - Vertices == rows.
  - Hyperedges == columns.
  - Weights == Communication volume for that edge  $\Rightarrow w_e \cdot (\# \text{ processors in edge} - 1)$ .
- Why Hypergraphs?
  - Models structurally non-symmetric systems.
  - Models communication costs — esp. important for non-homogeneous meshes.
  - Minimizes combined cost of application communication and data migration.



# Zoltan at Work



■ Data Migration  
Communication

■ Application  
Execution  
Communication

● 680k rows, 2.3M nonzeros



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# Conclusions

- RefMaxwell
  - Handles jumpy coefficients well.
  - Utilizes existing technology.
  - Scalability up to 24.5k procs!
- Zoltan & Hypergraph Repartitioning
  - Accurately models application and data migration costs.
  - Good results on test problems.
  - Future: ML's unstructured mesh applications.



# Future Directions

- TOPS-II: Interface w/ PETSc.
- Extreme Scalability.
- Adaptive methods for circuit problems.
- Improved multimaterial algorithms.