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Matrix Compression in the Method of Moments Code EIGER - Iterative Solver Accuracy and Parallel Efficiency

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Outline

- **EIGER Description**
- **Matrix compression**
- **Load- balancing (Matrix Fill)**
- **Results**
- **Conclusions / Future Work**

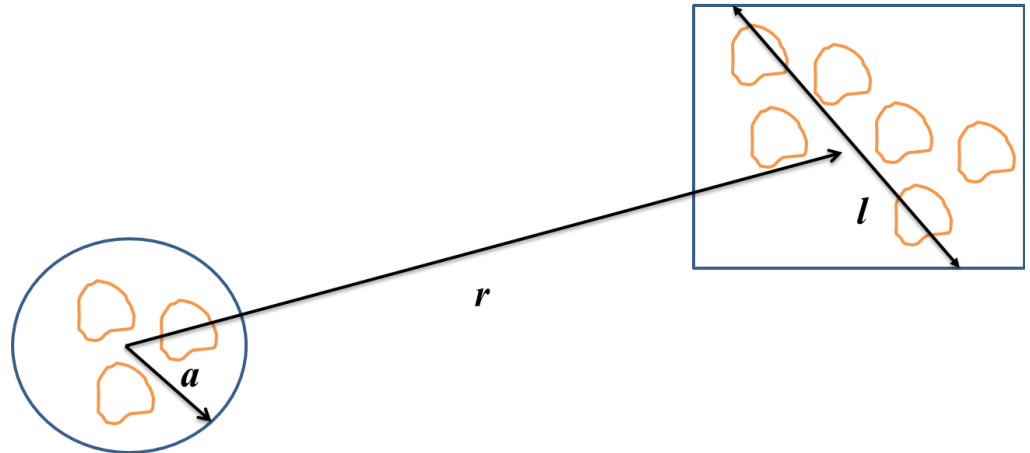
EIGER Description

- **Frequency-domain method of moments solution**
 - Steady state solution
 - F90 (95, 2003) code – Object Oriented Design
- **Boundary element formulation**
 - Mesh surfaces of parts – interface between regions
- **Normal formulation results in dense (fully populated) matrix**
 - Galerkin testing - Rao, Wilton, and Glisson bases functions
$$\bar{\bar{Z}} \bar{\bar{I}} = \bar{\bar{V}}$$
 - Simulations can be limited by available memory
(Entries are double precision complex)

Matrix Compression

- These are techniques that no longer store the full matrix but a lower rank version of the matrix.
- Based on work by Bucci and Franceschetti
 - “On the Degrees of Freedom of Scattered Fields” IEEE AP, July 1989

$$N_{dof} = \frac{4la}{r\lambda}$$



- **Fast Multipole Method (FMM)**
 - **Compression achieved through Green's function simplification:**
 - Factorization
 - Use of the addition theorem
 - Operator Diagonalization
 - Results in low-rank approximation of matrix blocks
 - $N \log(N)$ complexity

- **Adaptive Cross Approximation (ACA)**
 - **Compression achieved:**
 - Low-rank approximation of matrix blocks.
 - Done on the fly
 - Compressed matrix blocks never fully populated.
 - Since the process only operates on matrix blocks it is independent of Green's function simplification.
 - $N^{1.5}$ complexity

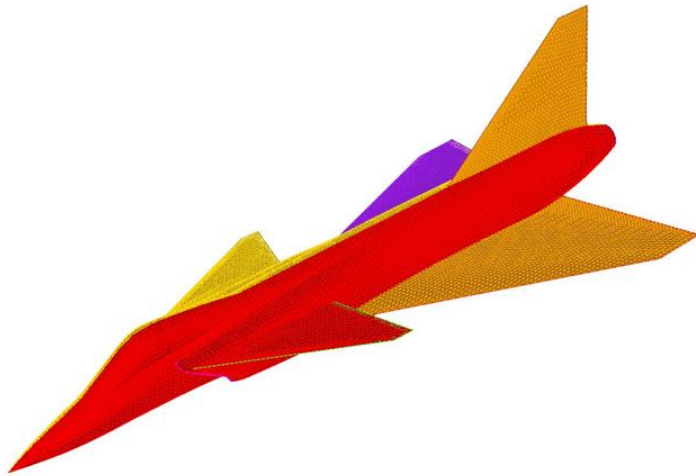
- **Multilevel direct solve methods**
 - University of Michigan, University of Colorado, and University of Texas at Austin

Matrix Compression Techniques

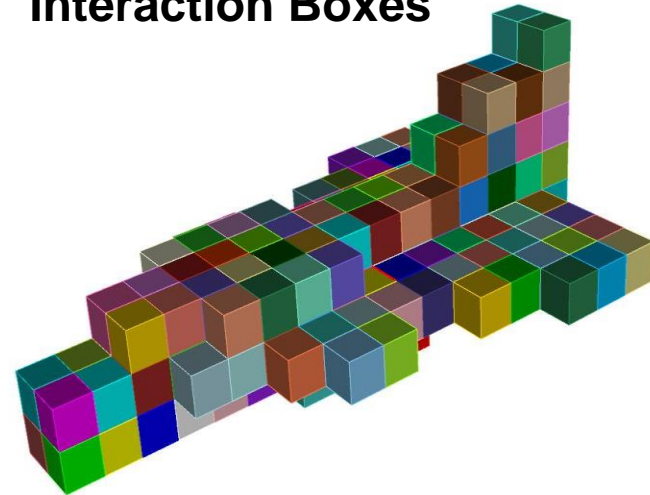
- Identification of all matrix blocks
 - Discretized object (meshed) is encased in a oct-tree structure

VFY 218

Meshed Object



Interaction Boxes



All compression techniques use this step in the solution process

ACA Matrix Compression

- Each box contains elements with current unknowns on the elements.
 - Can be compared to a 1-level fast multipole algorithm
- 2 boxes interact to form a matrix block.
- The distance between boxes, size of the boxes, and wavelength determine if a reduced or low-rank approximation can be used.
 - **Not all blocks can be compressed.**
 - **Compression criterion :**
 - Distance between the center of boxes $> 2 * (\text{box radius})$

ACA Matrix Compression

- The matrix $\bar{\bar{Z}}$ is given by:

$$\bar{\bar{Z}} = \sum_{j=1}^{MOM_blocks} Z_j^{mom} + \sum_{i=1}^{COM_blocks} \tilde{Z}_i^{com}$$

MOM_Blocks – *Moment method matrix blocks (full matrix blocks)*

COM_Blocks – *Compressed matrix blocks (low-rank approximation)*

ACA Matrix Compression

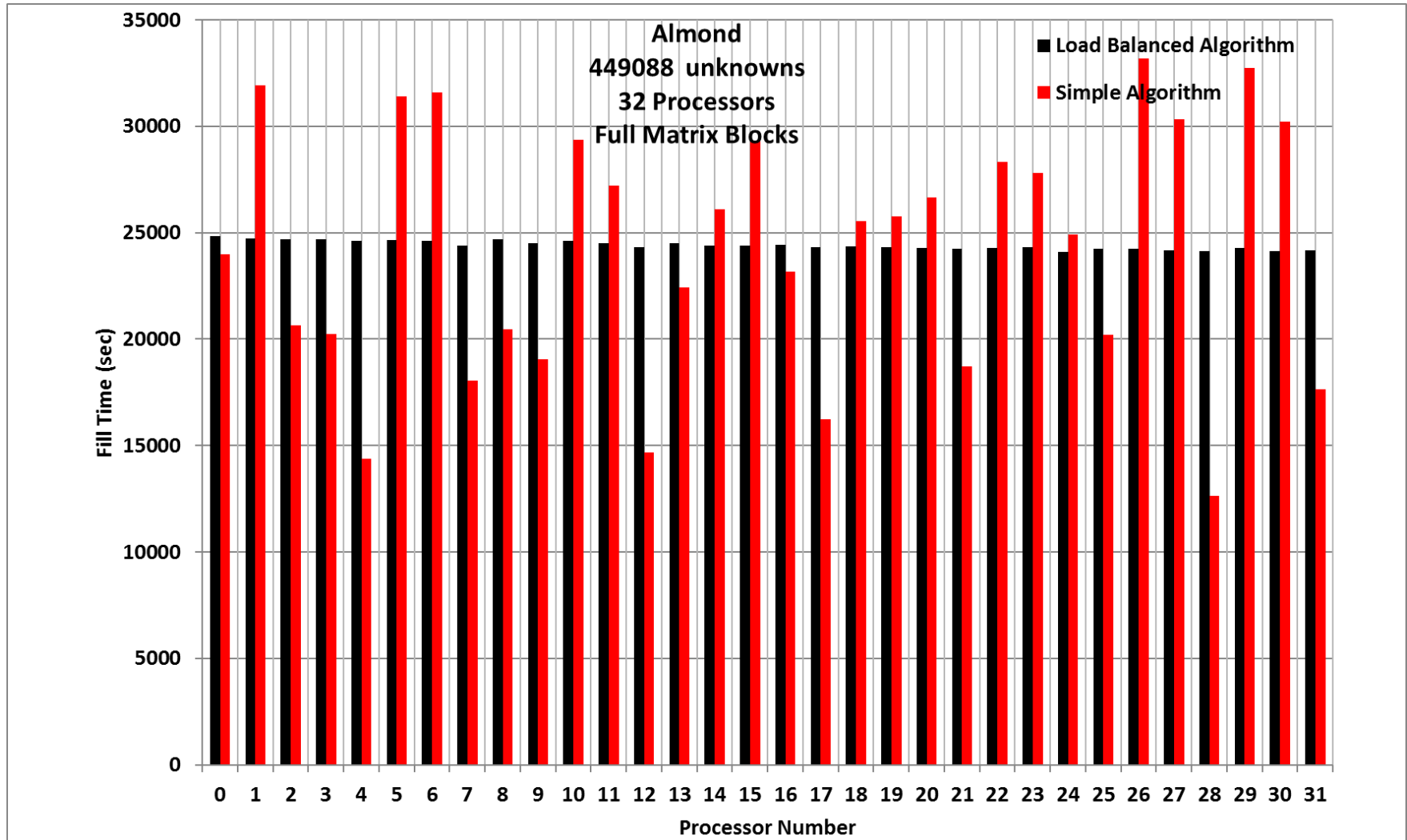
- Approximate matrix description:

$$\tilde{Z}^{m \times n} = U^{m \times r} V^{r \times n} = \sum_{i=1}^r u_i^{m \times 1} v_i^{1 \times n}$$

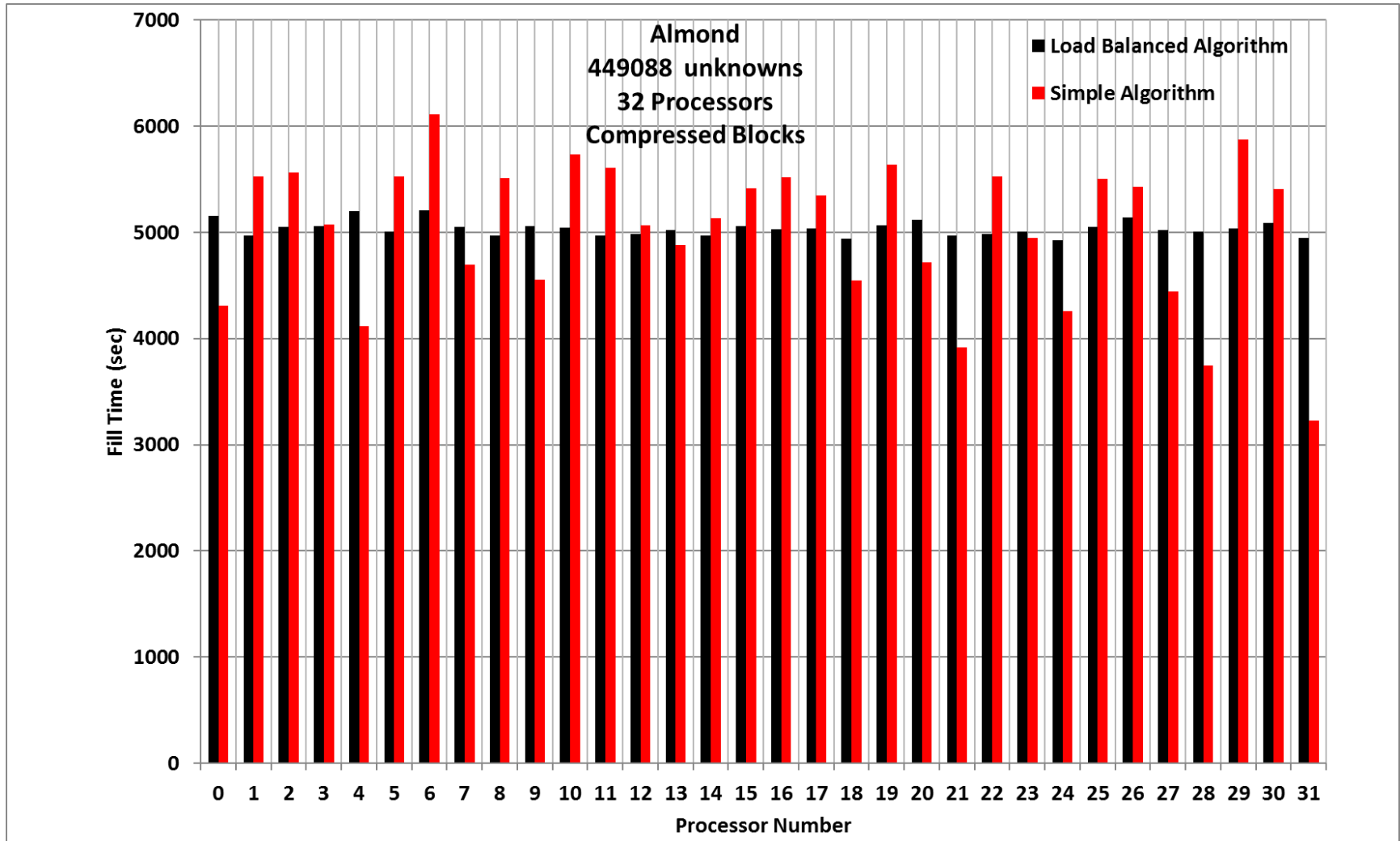
- The key step is the determination of the sub-matrices u and v .

- Each processor computes a subset of the total number of blocks
 - Both MOM_Blocks and COM_Blocks
- Initial decomposition of blocks done via block numbering
 - Neglects the work per block
- Improved decomposition of blocks performed
 - Sorted with regards to matrix block size
 - Known for full matrix blocks
 - Estimated using “Degrees of Freedom of the Scattered Field”

Parallel Computation of Matrix Blocks



Parallel Computation of Matrix Blocks



Solution of the Compressed System

- The matrix equation to be solved is :

$$\bar{\bar{Z}} \bar{\bar{I}} = \bar{\bar{V}}$$

- The matrix is not completely available but is stored as:

$$\bar{\bar{Z}} = \sum_{j=1}^{MOM_blocks} Z_j^{mom} + \sum_{i=1}^{COM_blocks} \tilde{Z}_i^{com}$$

- Therefore a iterative solution approach needs to be used.
 - Generalized Minimum residual method(GMRES)
 - Saad and Schultz 1986
 - Transpose Free Quasi Minimum Residual (TFQMR)
 - Freund 1993
 - Both exercised and require matrix – vector products

Solution of the Compressed System Sandia National Laboratories

- **Low – rank approximation of the compressed sub matrices controlled by a parameter. (eps)**

- **Iterative solution controlled by chosen residual**
 - **For GMRES (number of vectors for restart)**

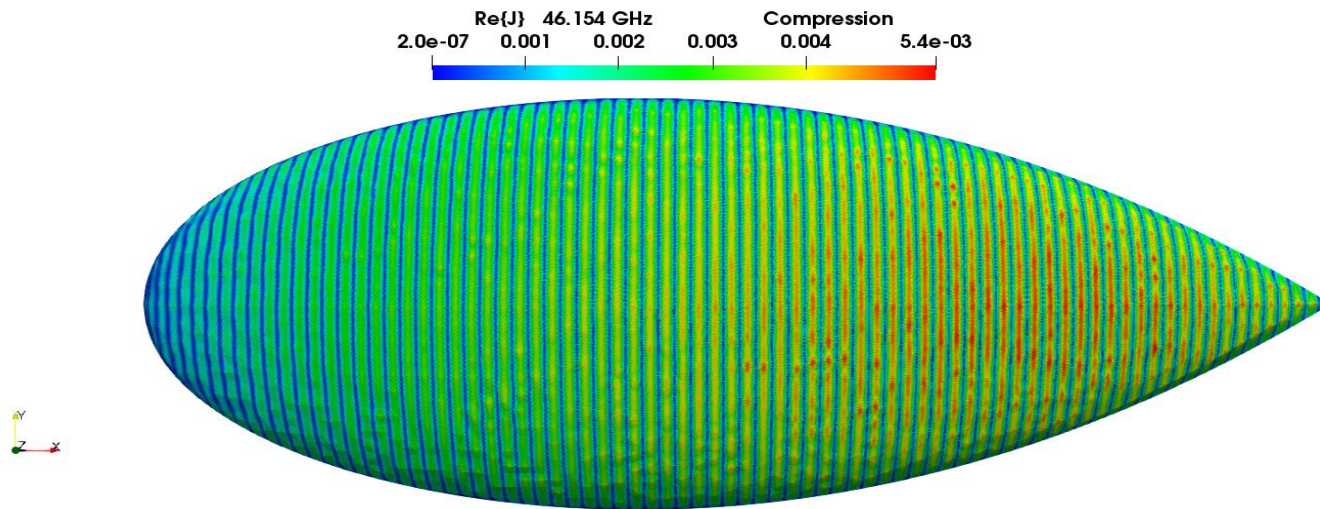
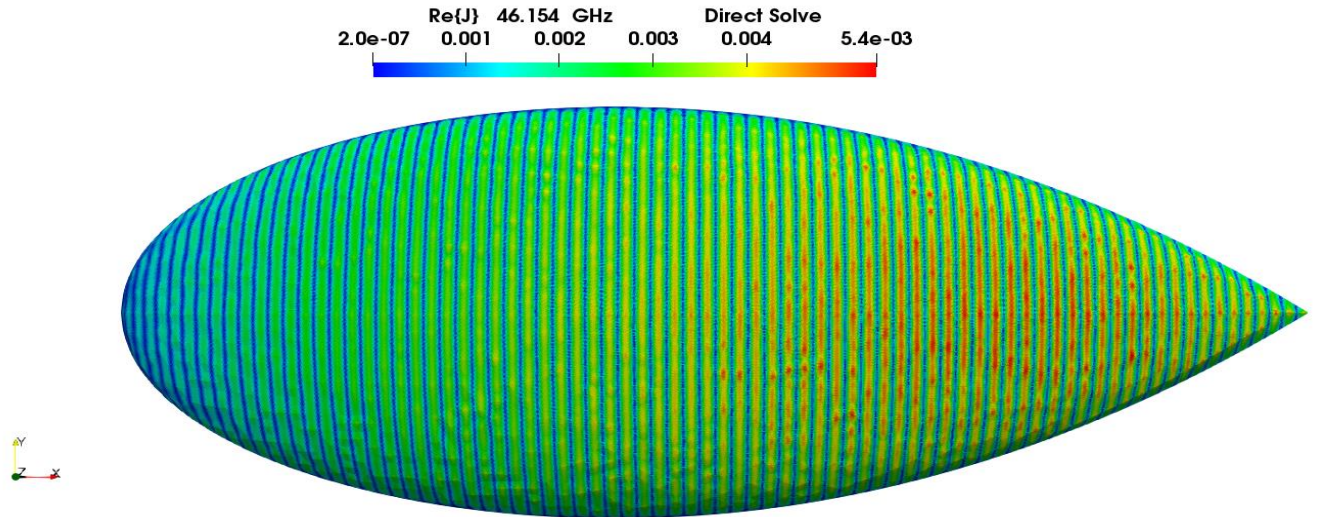
Test Object Almond

- **499088 Unknowns**
 - **Grid for 50 GHz**

Solve	# processors	Memory (Gbytes)	Solve Time (sec)
Direct	1980	3226	17992
Compression	48	322 (90%)	891

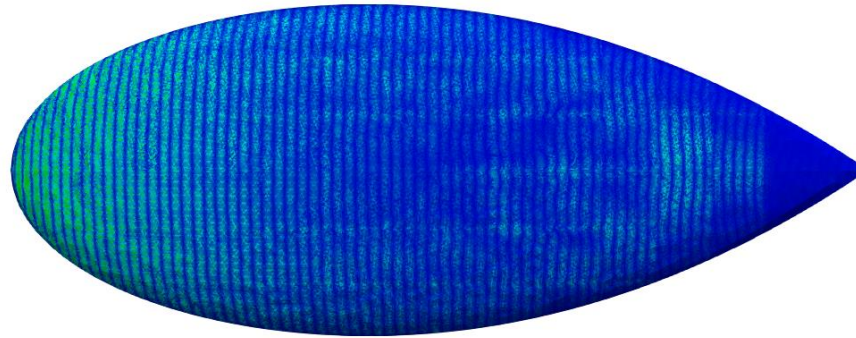
Eps 4e-03, Solver tolerance 1.e-06, GMRES (40)

Surface Currents Comparison

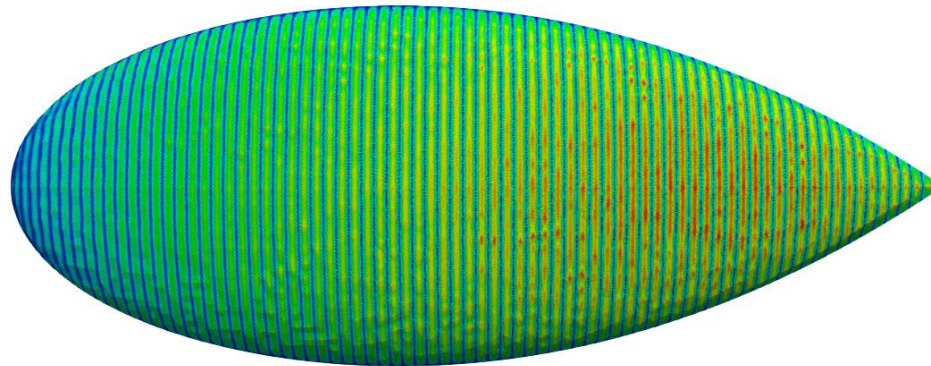


Surface Currents Comparison

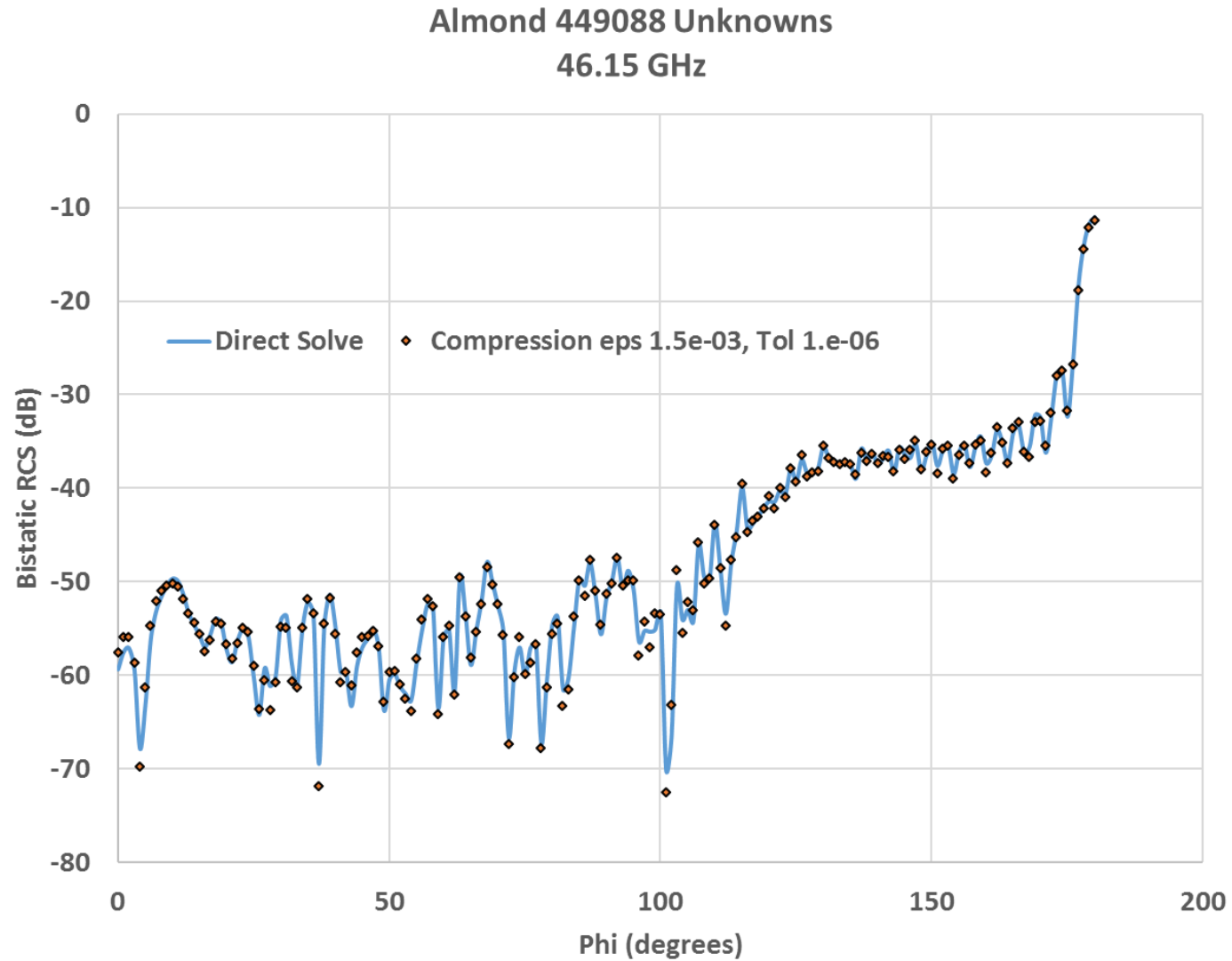
abs (Re {J_{ju} - J_{comp}}) 46.16 GHz Magnitude
2.0e-07 0.001 0.002 0.003 0.004 5.4e-03



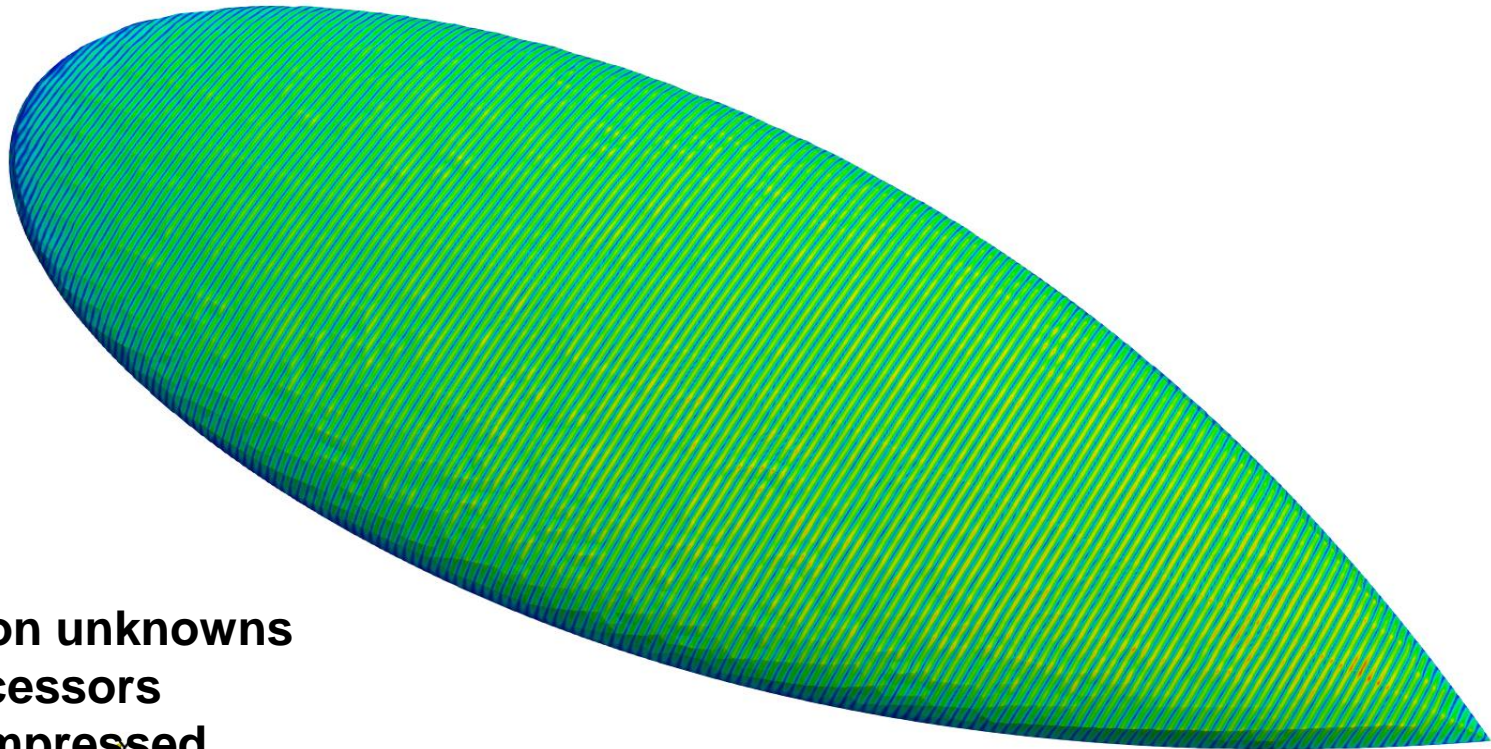
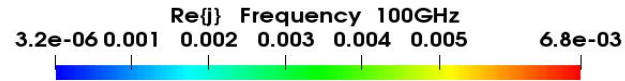
Re{J} 46.15 GHz
2.0e-07 0.001 0.002 0.003 0.004 5.4e-03



Far Field Comparison



Surface Currents – Problem 2



1.8 million unknowns
150 Processors
98% Compressed
Eps $1.5e-03$
Solver Tol $1e-06$
Solution Time 685 Sec

Conclusions / Future Work

- Presented work using matrix compression (via ACA).
- Increase efficiency and robustness
 - Memory usage
 - Fill time
 - Matrix-vector product
- Collaboration with Solver Team (Trilinos)
 - Preconditioning – necessary for the EXTERNAL / INTERNAL problem
 - Block methods (Multiple excitation / directions)
 - MPI + threading
- Continue to extend solution frequency applicability
 - Alternative methods of compression