



Unstructured Grid Technology for CFD: Flow Solver Perspectives

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Overview

- Unstructured mesh technology has come of age (finally)
 - Evidence from Drag Prediction Workshops
- Efficient solvers required for steady-state and time-implicit problems
 - (similar to structured meshes)
- Grid resolution and convergence
- AMR
- Sensitivity Analysis
- Higher-Order Methods

Unstructured Mesh Technology

- Current evidence suggests
 - USG accuracy comparable to structured grid accuracy on grid of similar resolution
 - Current best practice:
 - Prismatic elements in boundary layer regions
 - Tetrahedral elements in inviscid regions
 - Shock alignment for hypersonics may require hexahedral elements (Candler, Gnoffo)
 - Cell vs. Node schemes
 - » Tradeoffs still not fully understood

Unstructured Mesh Technology

- Current evidence suggests
 - USG efficiency similar to structured meshes
 - Factor 2 to 3 more costly per iteration
 - Better scalability: Homogeneous data-structures
 - Assuming similarly capable solution strategies
 - Implicit, Multigrid
 - Becomes more critical for finer meshes
 - » Steady-state
 - » Low reduced frequency problems (aeroelastics)

NSU3D Solver

- Governing Equations: Reynolds Averaged Navier-Stokes Equations
 - Conservation of Mass, Momentum and Energy
 - Turbulence models
 - Spalart-Allmaras 1 equation model
 - K-omega 2 equation model
- Vertex-Based Discretization
 - 2nd order upwind finite-volume scheme
 - 6 or 7 variables per grid point
 - Flow equations fully coupled (5x5)
 - Turbulence equation uncoupled (1 or 2X2)

Spatial Discretization

- Mixed Element Meshes
 - Tetrahedra, Prisms, Pyramids, Hexahedra
- Control Volume Based on Median Duals
 - Fluxes based on edges

* $\mathbf{F}_{ik} = f(\mathbf{u}_{\text{left}}, \mathbf{u}_{\text{right}})$

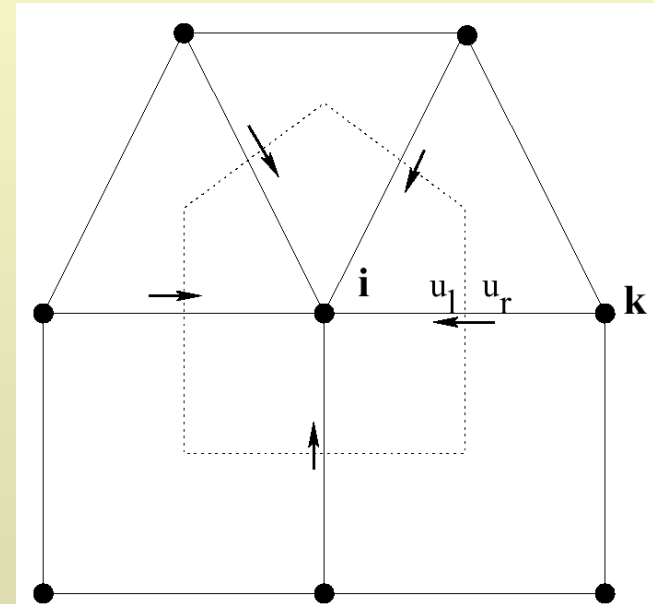
* $\mathbf{u}_{\text{left}} = \mathbf{u}_i, \mathbf{u}_{\text{right}} = \mathbf{u}_k$: *1st order accurate*

* $\mathbf{u}_{\text{left}} = \mathbf{u}_i + \frac{1}{2} \nabla \mathbf{u}_i \cdot \mathbf{r}_{ik}$

* $\mathbf{u}_{\text{right}} = \mathbf{u}_k + \frac{1}{2} \nabla \mathbf{u}_k \cdot \mathbf{r}_{ki}$: *2nd order accurate*

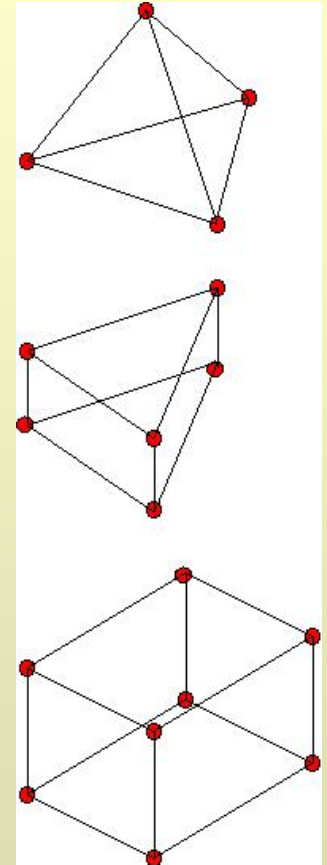
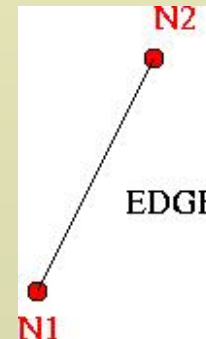
– * ∇u_i *evaluated as contour integral around CV*

- Various reconstruction options
 - Least squares, Green-Gauss,
 - biharmonic differences (matrix artificial dissipation)



Mixed-Element Discretizations

- Edge-based data structure
 - Building block for all element types
 - Reduces memory requirements
 - Minimizes indirect addressing / gather-scatter
 - Graph of grid = Discretization stencil
 - Implications for solvers, Partitioners



Solution Methodology

- To solve $R(w) = 0$ (steady or unsteady residual)

- Newton's method:
$$\left[\frac{\partial R}{\partial w} \right] \Delta w^{n+1} = -R(w^n)$$
- Requires storage/inversion of Jacobian $\left[\frac{\partial R}{\partial w} \right]$ (too big for 2nd order scheme)
- Replace with 1st order Jacobian
 - Stored as block Diagonals [D] (for each vertex)
and off-diagonals [O] (2 for each edge)
- Use block Jacobi or Gauss-Seidel to invert Jacobian at each Newton iteration using subiteration k:

$$[D] \Delta w^{k+1} = -R(w^n) - [O] \Delta w^k$$

Solution Methodology

- Corresponds to linear Jacobi/Gauss-Seidel in many unstructured mesh solvers

$$[D]\Delta w^{k+1} = -R(w^n) - [O]\Delta w^k$$

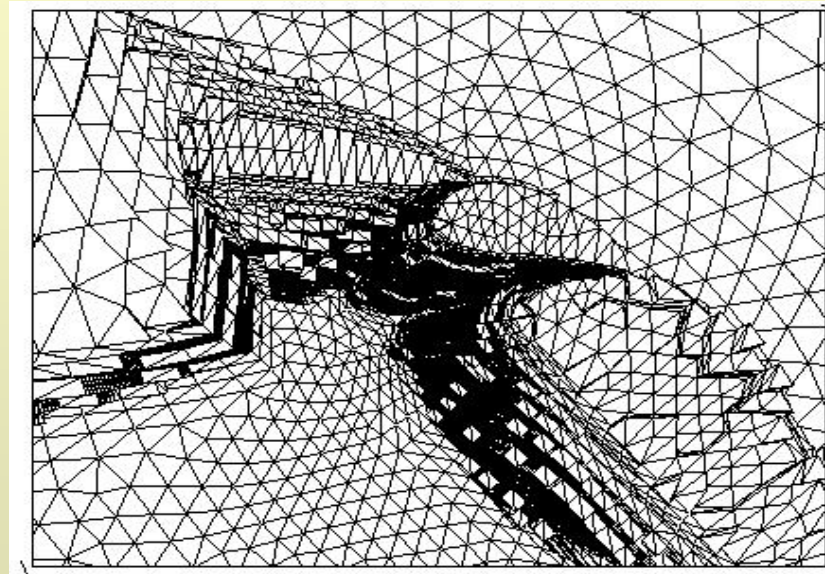
- Alternately, replace Jacobian simply by [D] (i.e. drop [O] terms) (Point implicit)

$$[D]\Delta w^{n+1} = -R(w^n)$$

- Non-linear residual must now be updated at every iteration (no subiterations)
- Corresponds to non-linear Jacobi/Gauss-Seidel
- Converges at same rate as linear scheme (in index k) in the absence of strong non-linearities
 - Non-linear scheme requires ~ 150 words /cv
 - Linear schemes require ~500 words/cv : [O] ~ 350 words/cv
 - But non-linear scheme requires more cpu time because of more non-linear evaluations
- NSU3D production code employs non-linear solver

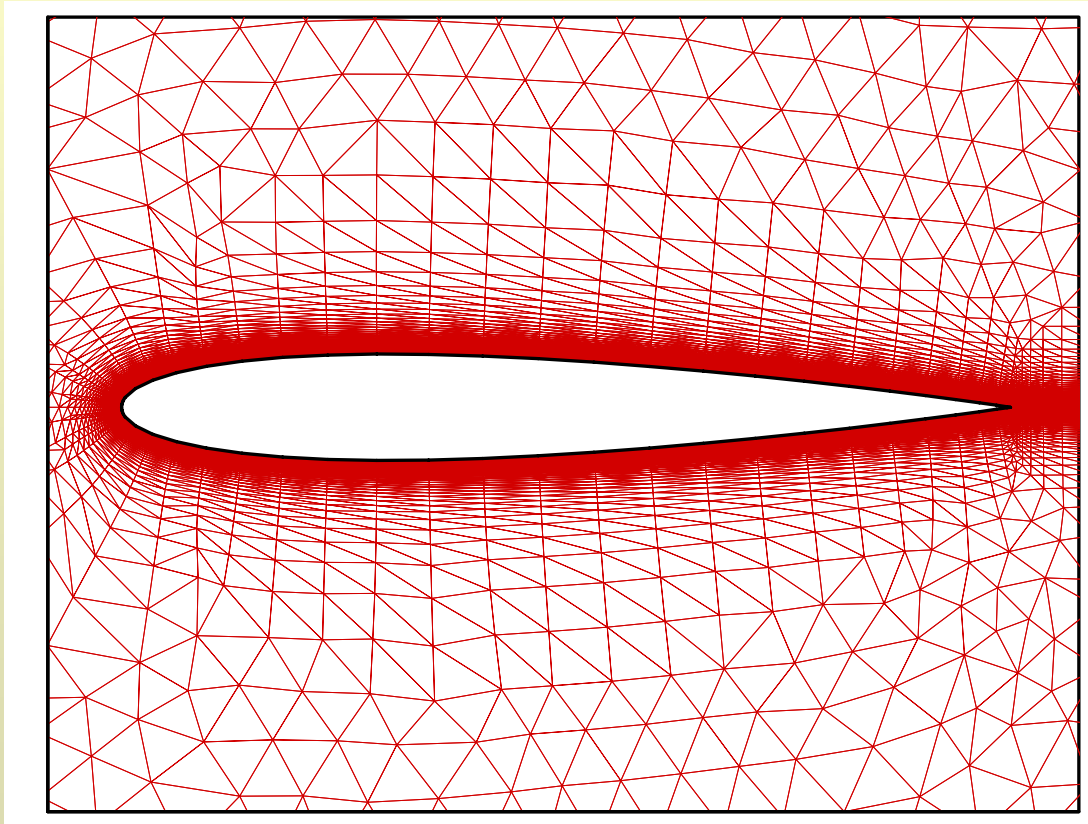
Anisotropy Induced Stiffness

- Convergence rates for RANS (viscous) problems much slower than inviscid flows
 - Mainly due to grid stretching
 - Thin boundary and wake regions
 - Mixed element (prism-tet) grids
- Use directional solver to relieve stiffness
 - Line solver in anisotropic regions



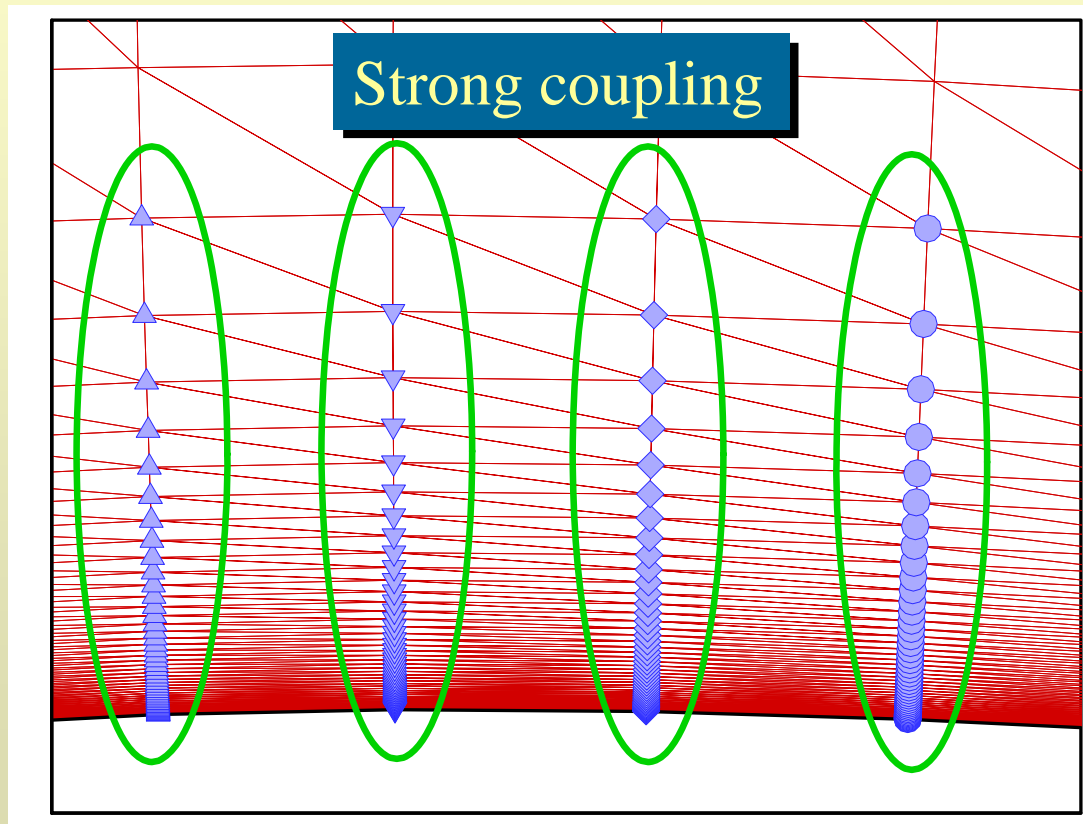
Method of Solution

- Line-implicit solver



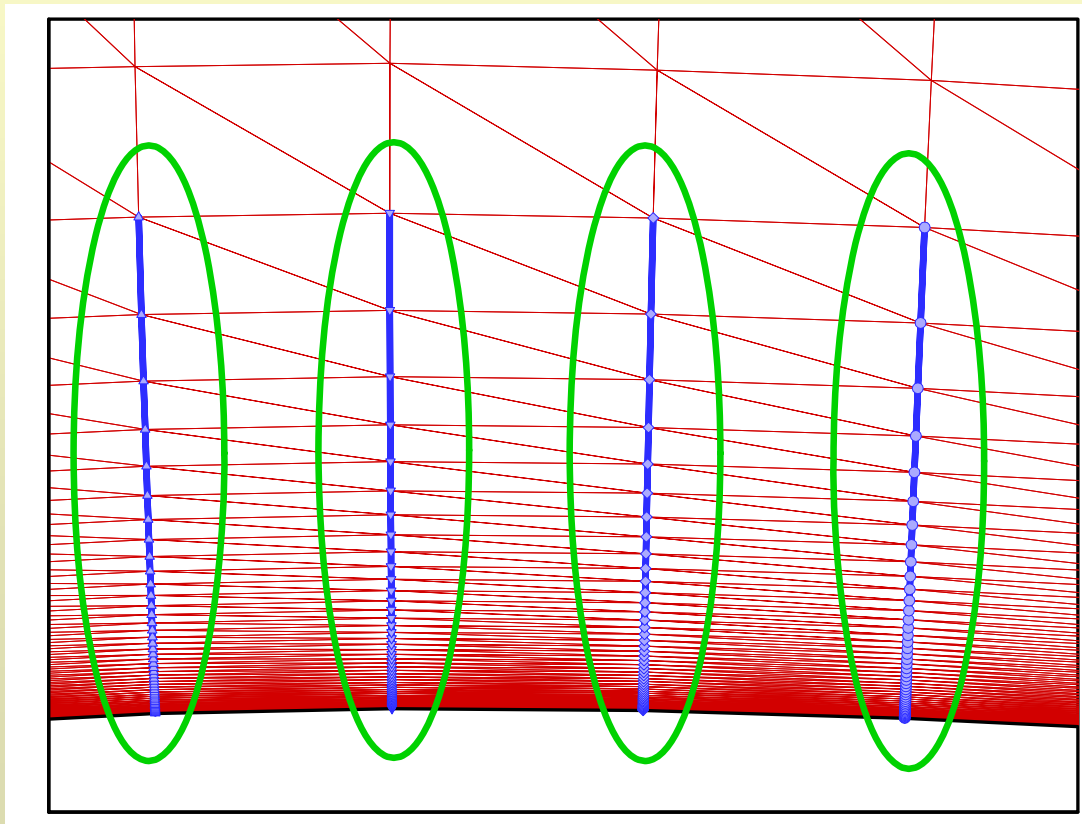
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Method of Solution

- Line-implicit solver

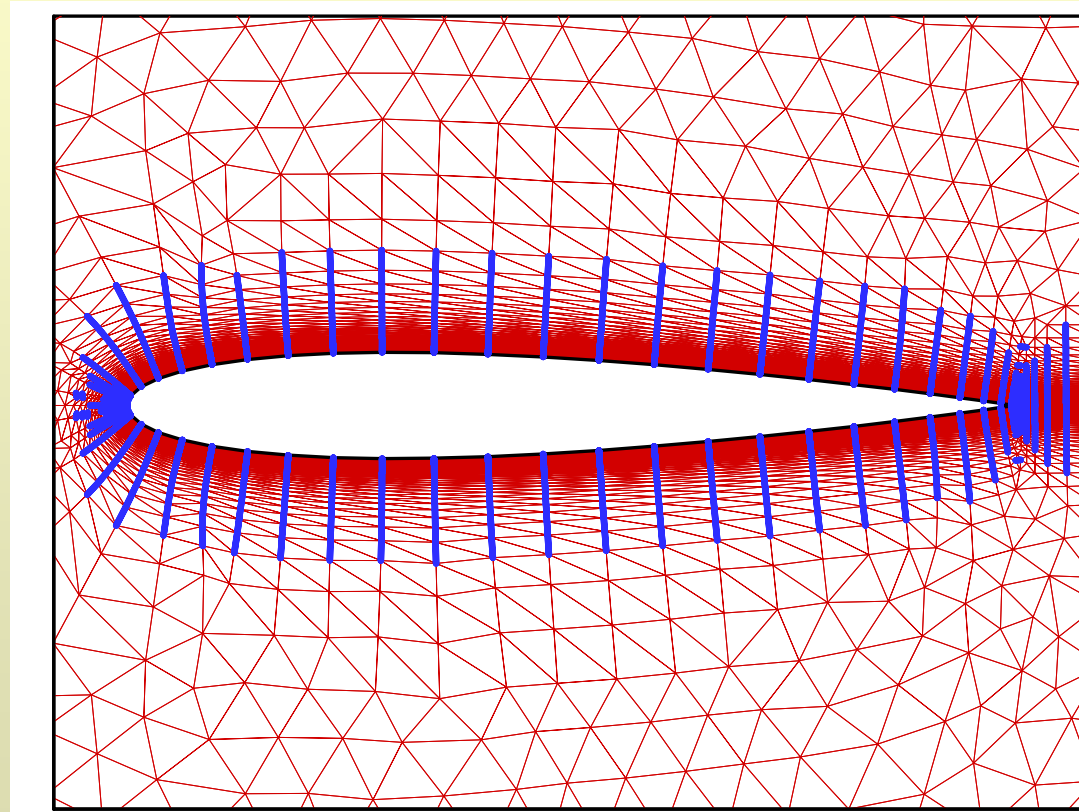


Method of Solution

- Line-implicit solver

$$[D]\Delta w^{n+1} = -R(w^n)$$

- [D] now represents point Jacobians plus off-diagonals corresponding to edges in lines (~25 extra words/cv)**
- Invert [D] using block tridiagonal algorithm**
- Analogous approach possible for linear sub-iterative solver
- Gauss-Seidel extensions
- **: Non-linear line Jacobi used in production code

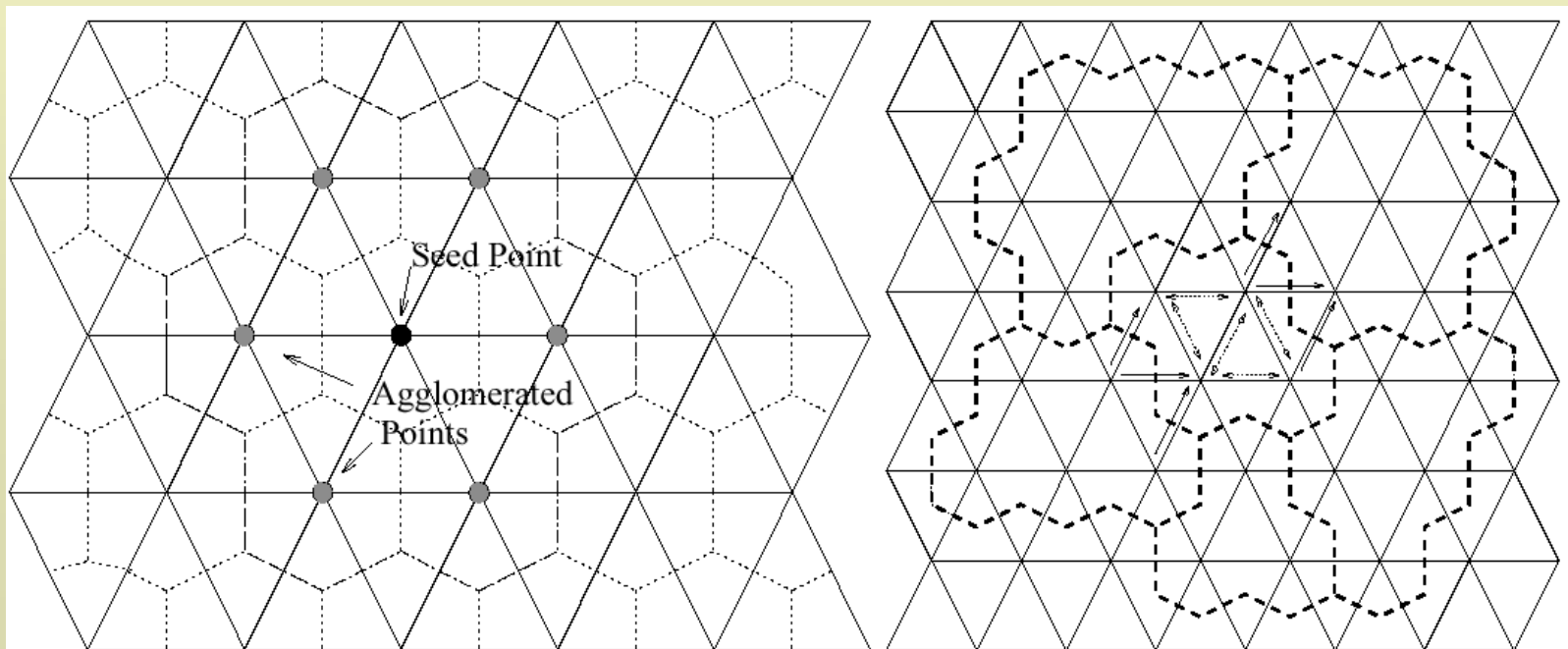


Method of Solution

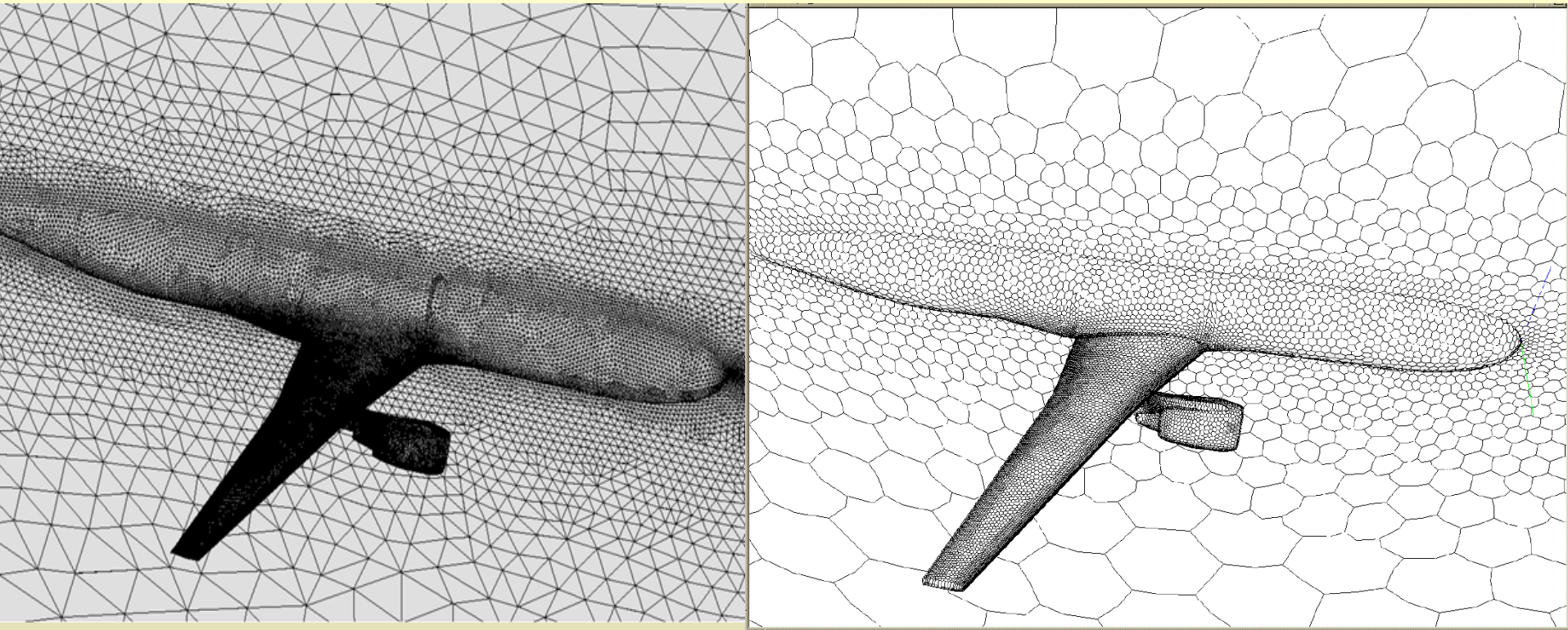
- Line Jacobi/Gauss-Seidel solver effective for time-dependent problems with small time step
 - Convergence rates independent of grid anisotropy
- For large time steps or steady-state problems require more global solver
- Use line solver as driver on each level of agglomeration multigrid algorithm

Agglomeration Multigrid

- Agglomeration Multigrid solvers for unstructured meshes
 - Coarse level meshes constructed by agglomerating fine grid cells/equations

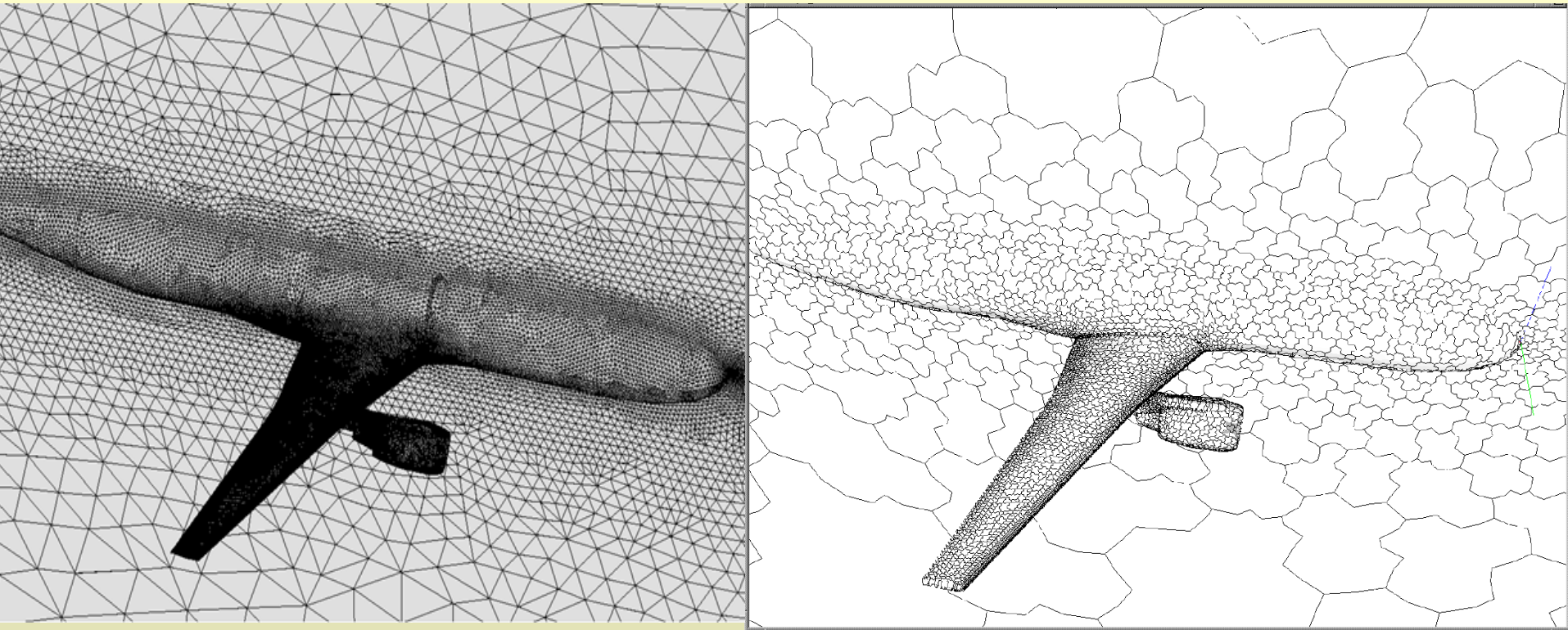


Agglomeration Multigrid



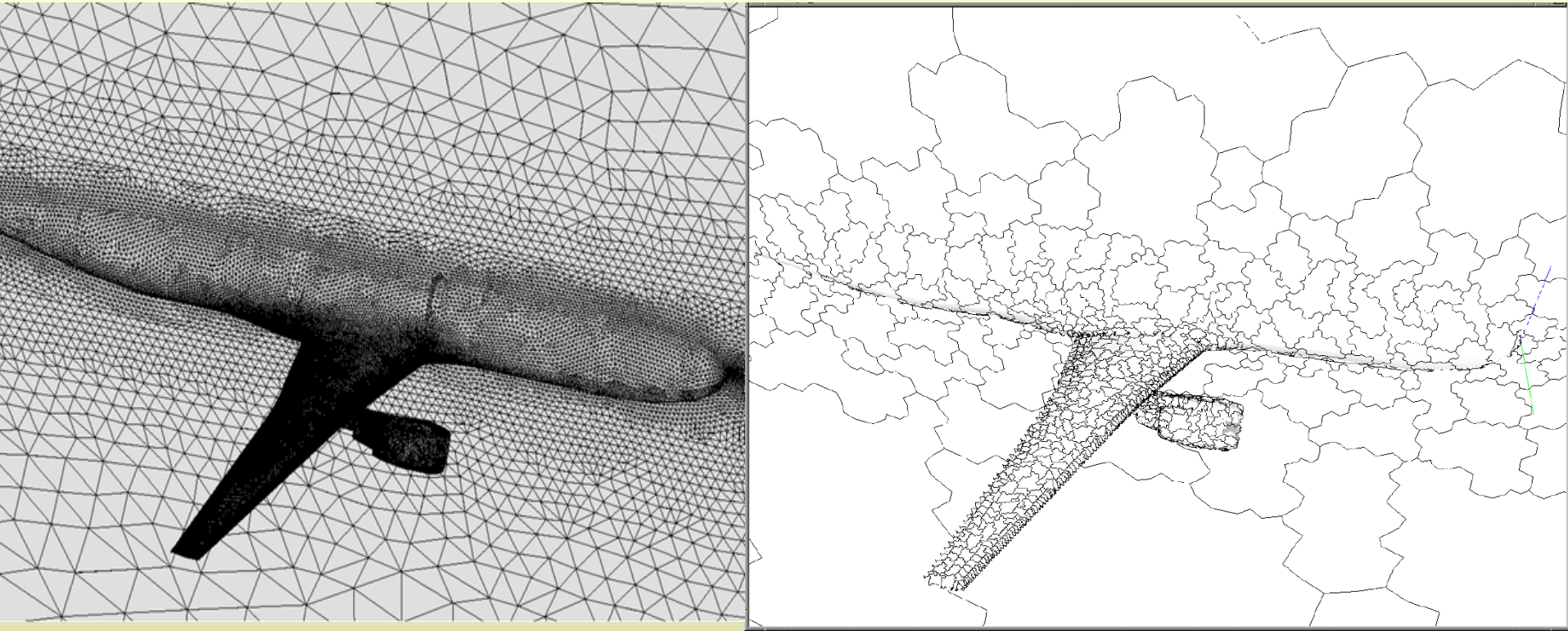
- Automated Graph-Based Coarsening Algorithm
- Coarse Levels are Graphs
- Coarse Level Operator by Galerkin Projection
- Grid independent convergence rates (order of magnitude improvement)

Agglomeration Multigrid



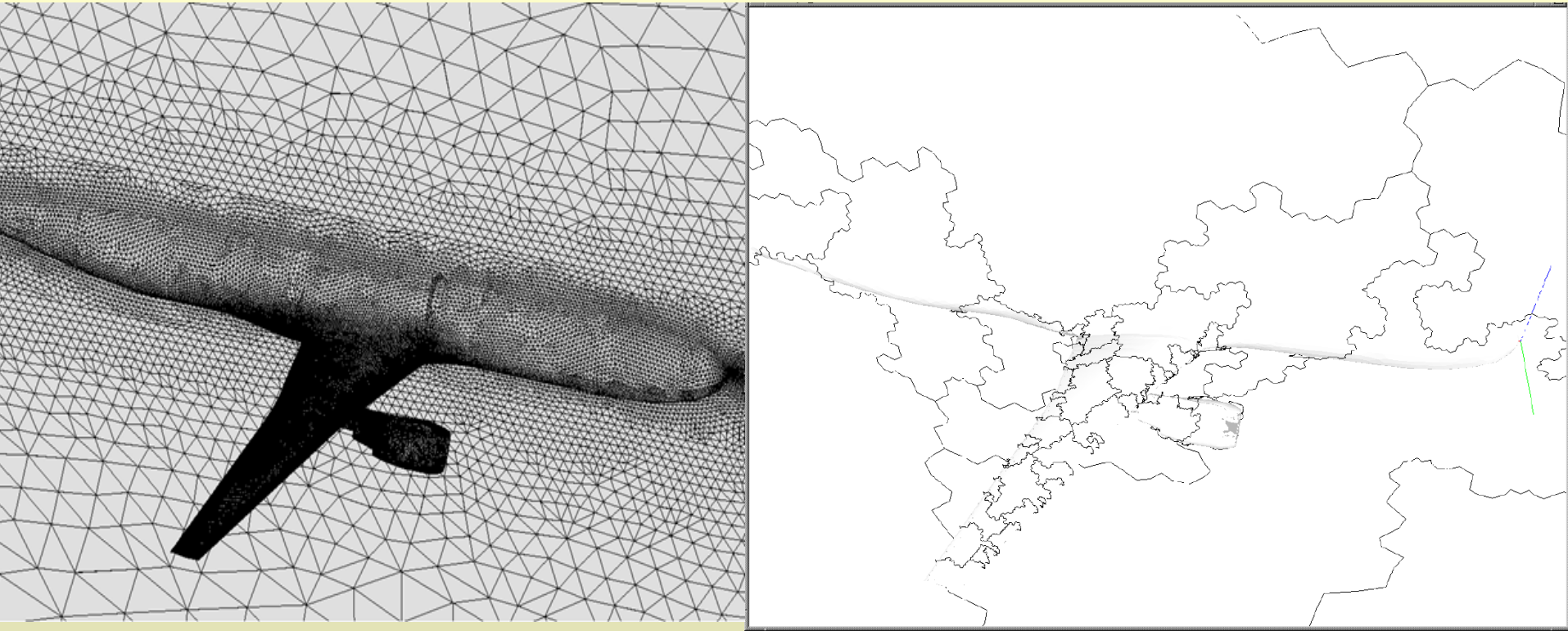
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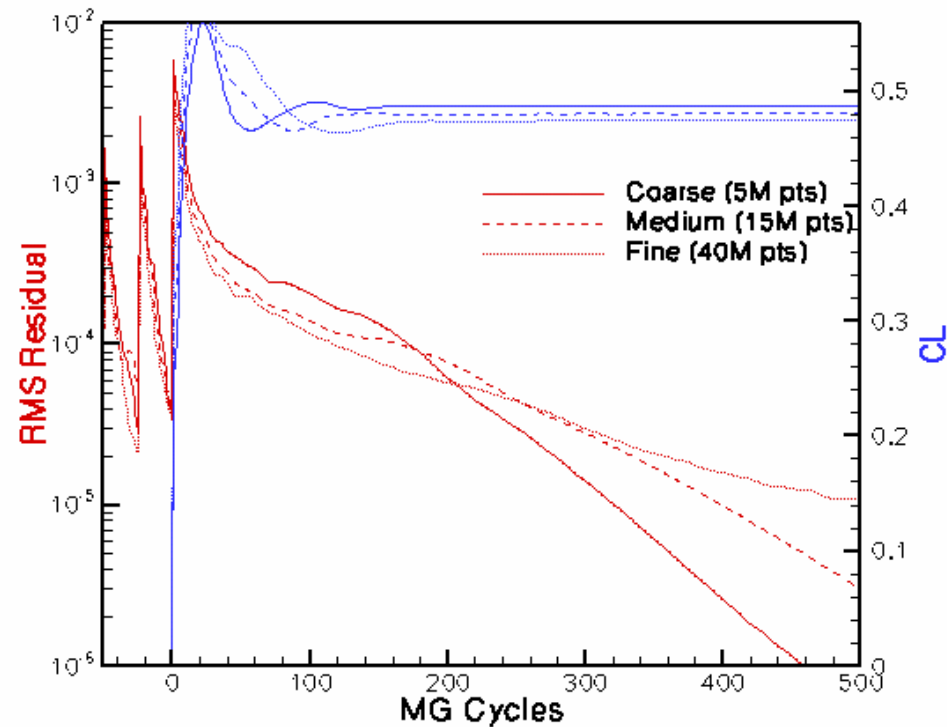
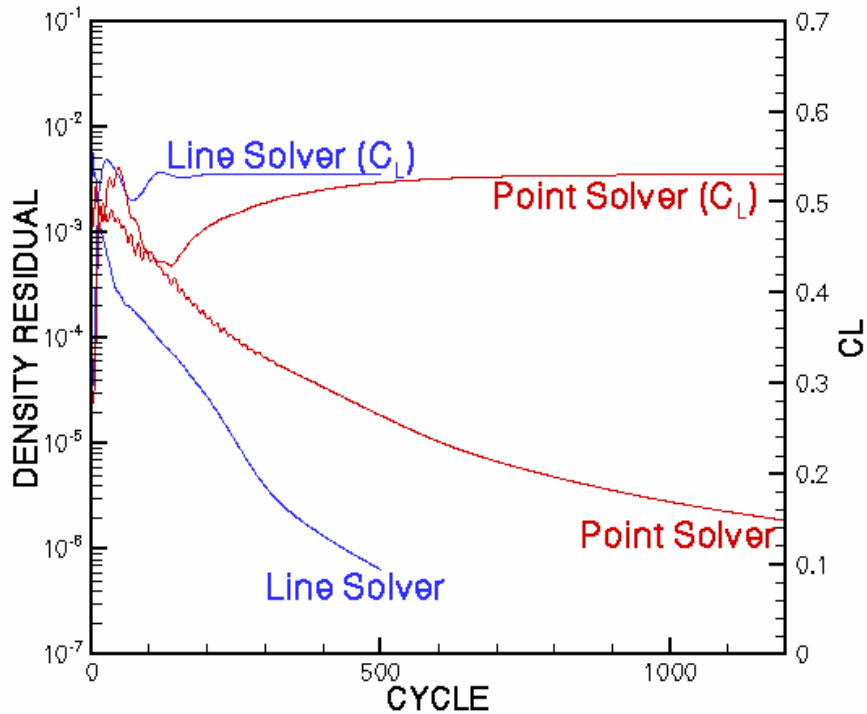
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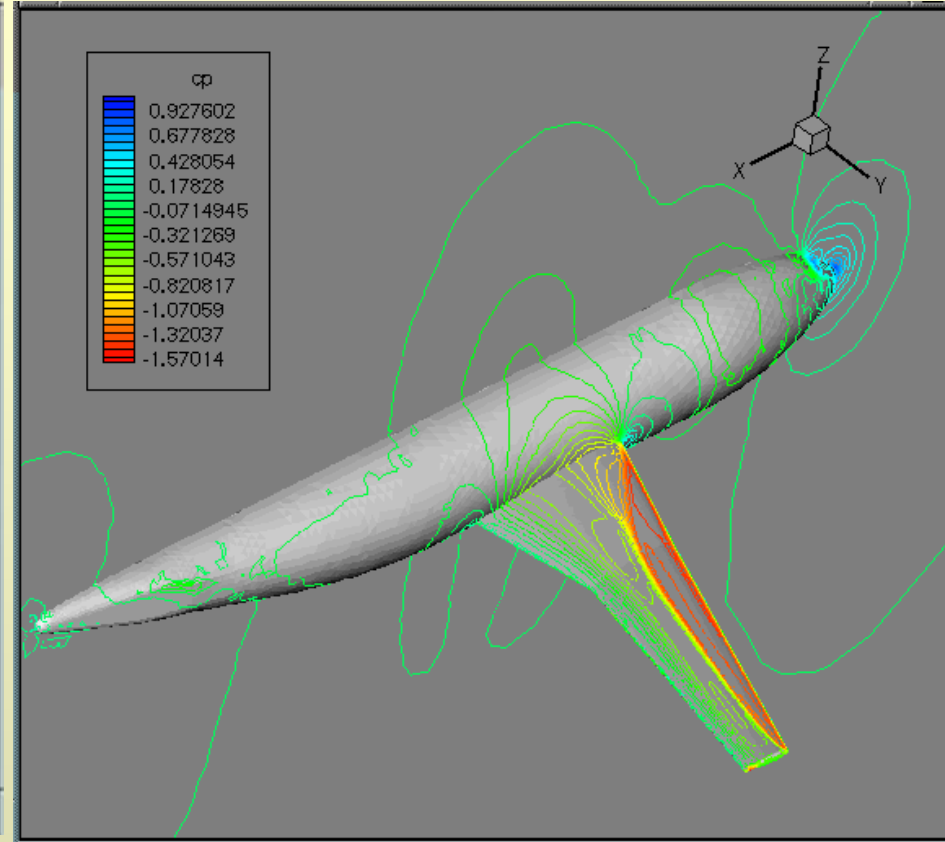
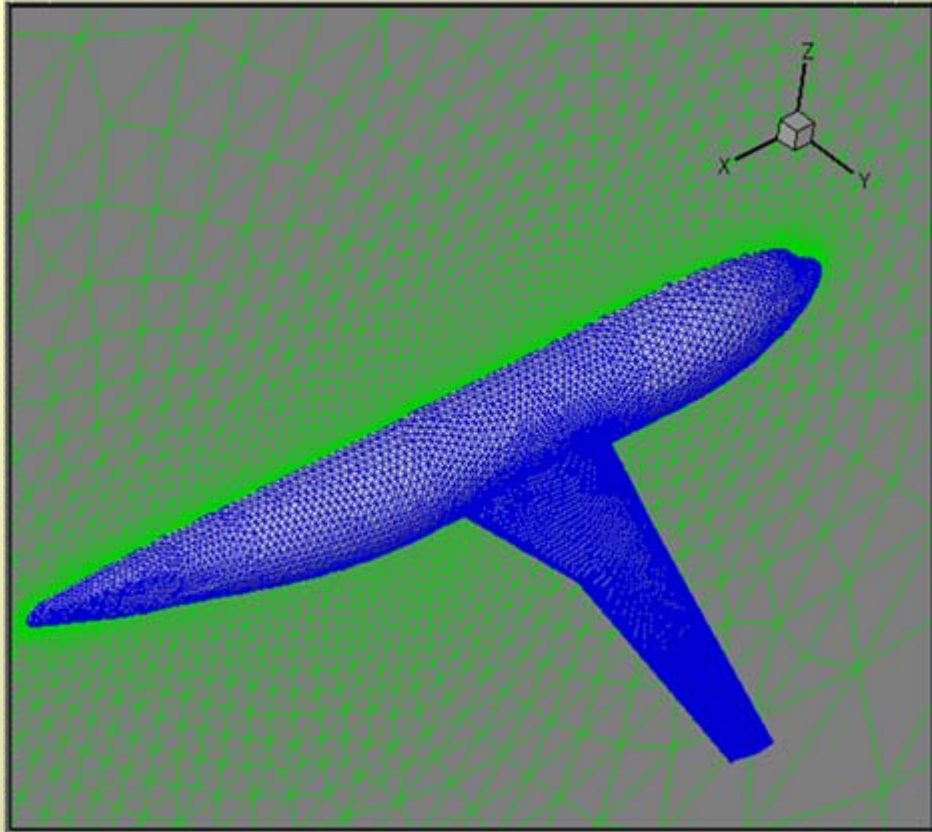
Line Solver Multigrid Convergence



Line solver convergence
insensitive to grid
stretching

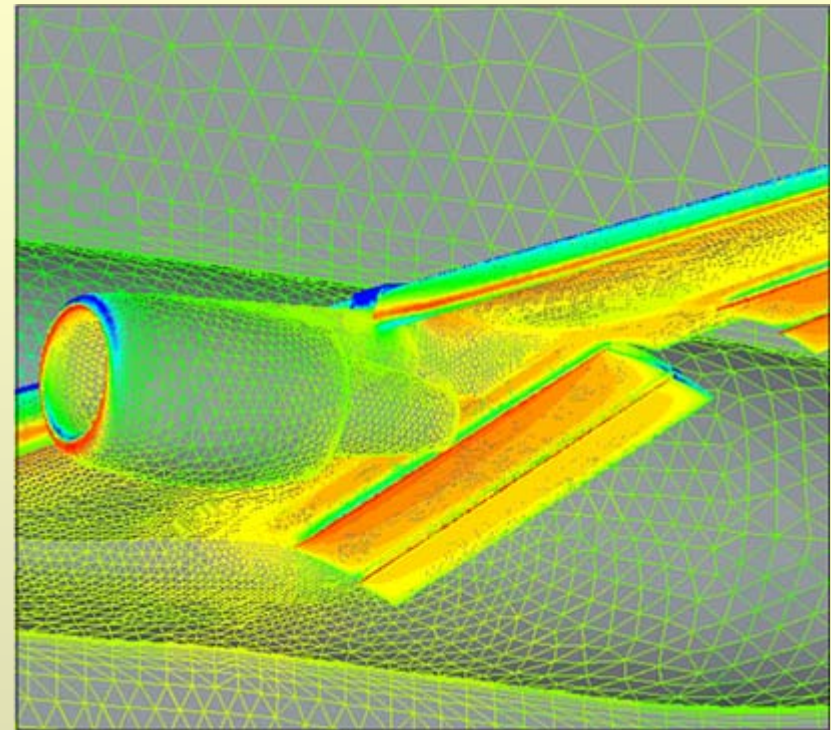
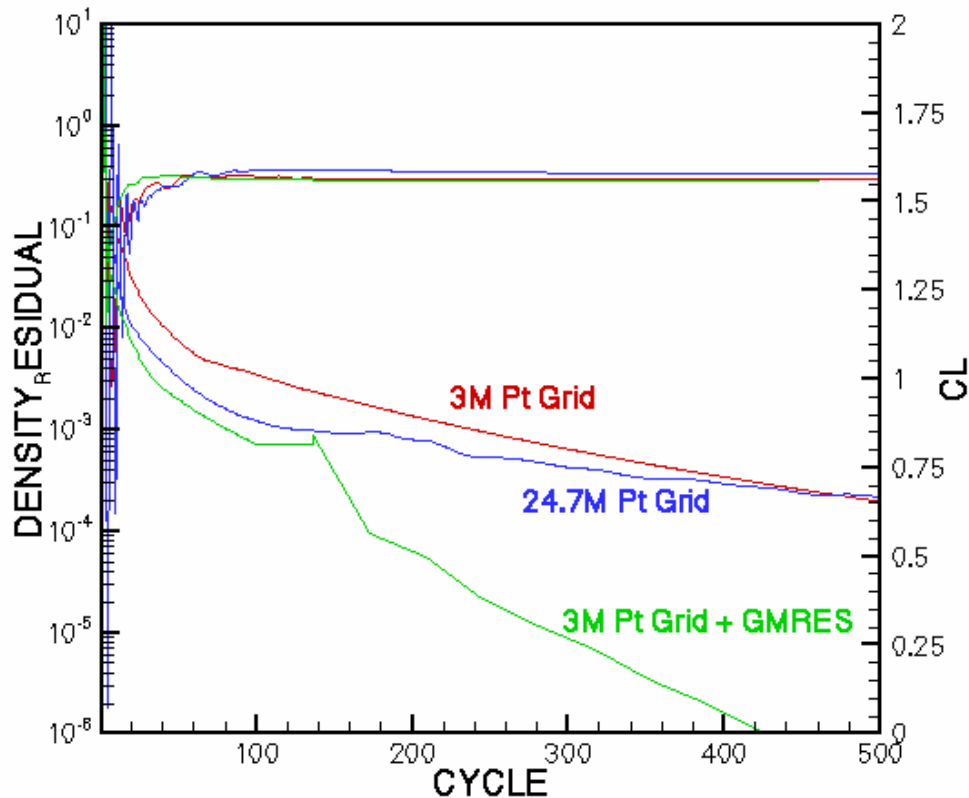
Multigrid convergence
insensitive to grid
resolution

NSU3D TEST CASE



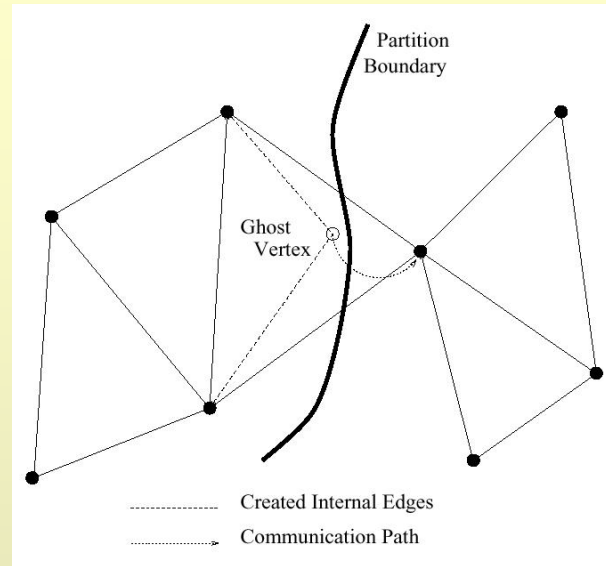
- Wing-Body Configuration
- 3 to 72 million grid points
- Transonic Flow
- Mach=0.75, Incidence = 0 degrees, Reynolds number=3,000,000

(Multigrid) Preconditioned Newton Krylov



- Mesh independent property of Multigrid
- GMRES effective (in asymptotic range) but requires extra memory

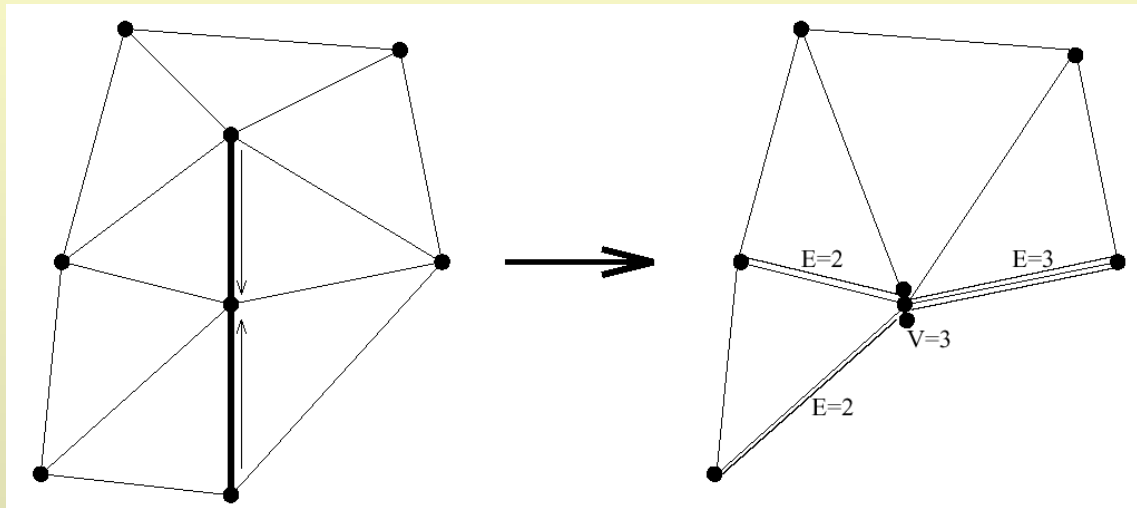
Parallelization through Domain Decomposition



- Intersected edges resolved by ghost vertices
- Generates communication between original and ghost vertex
 - Handled using MPI and/or OpenMP (Hybrid implementation)
 - Local reordering within partition for cache-locality

Partitioning

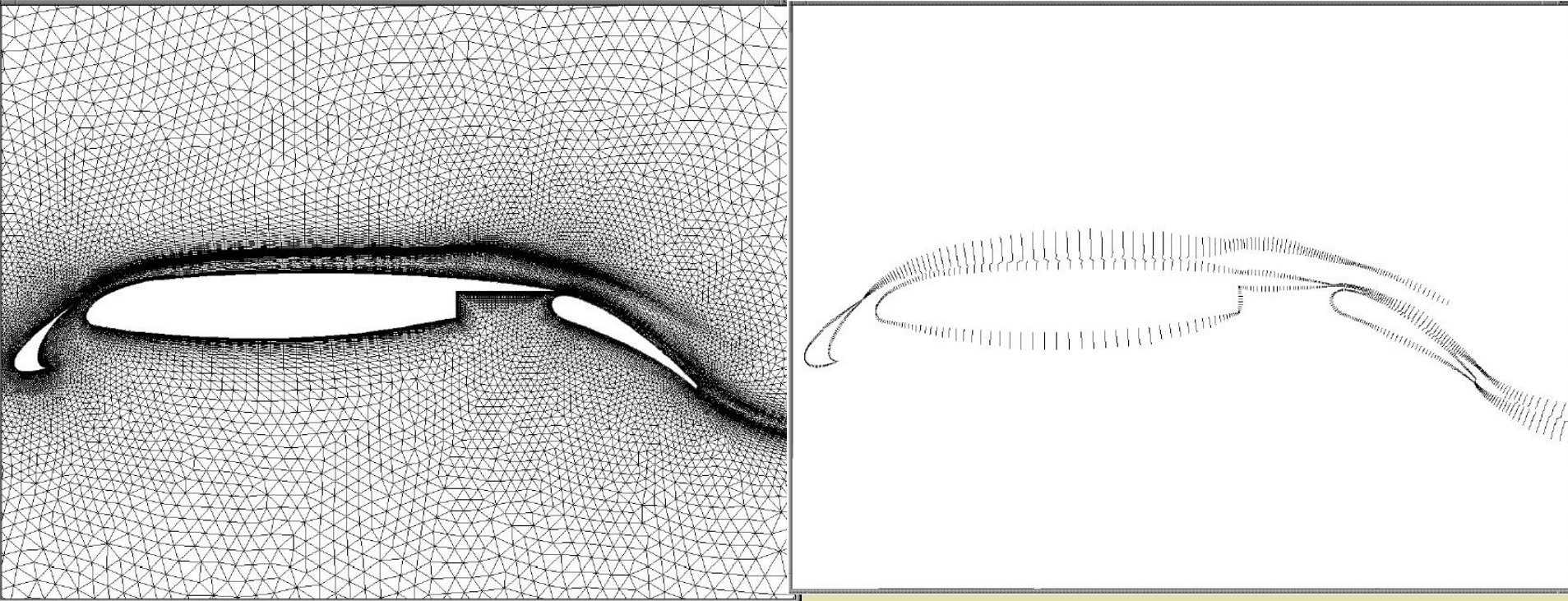
- (Block) Tridiagonal Lines solver inherently sequential
- Contract graph along implicit lines
- Weight edges and vertices



- Partition contracted graph
- Decontract graph
 - Guaranteed lines never broken
 - Possible small increase in imbalance/cut edges

Partitioning Example

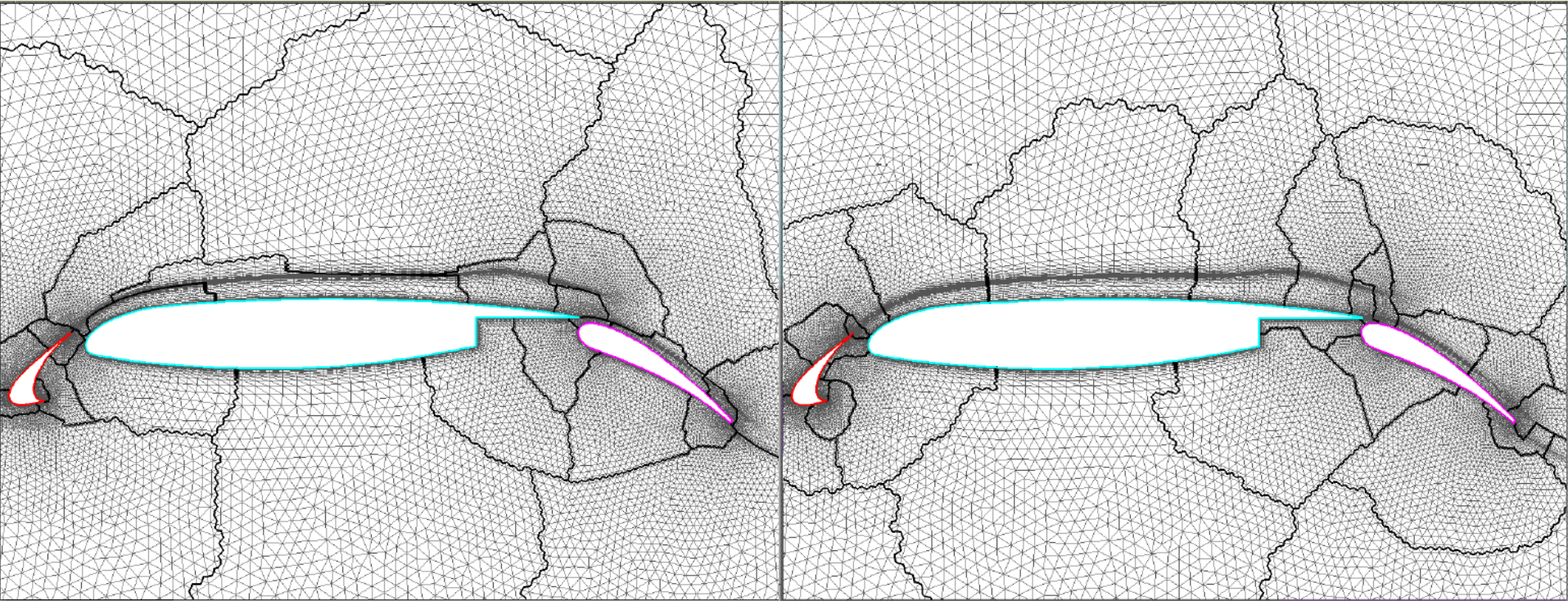
- 32-way partition of 30,562 point 2D grid



- Unweighted partition: 2.6% edges cut, 2.7% lines cut
- Weighted partition: 3.2% edges cut, 0% lines cut

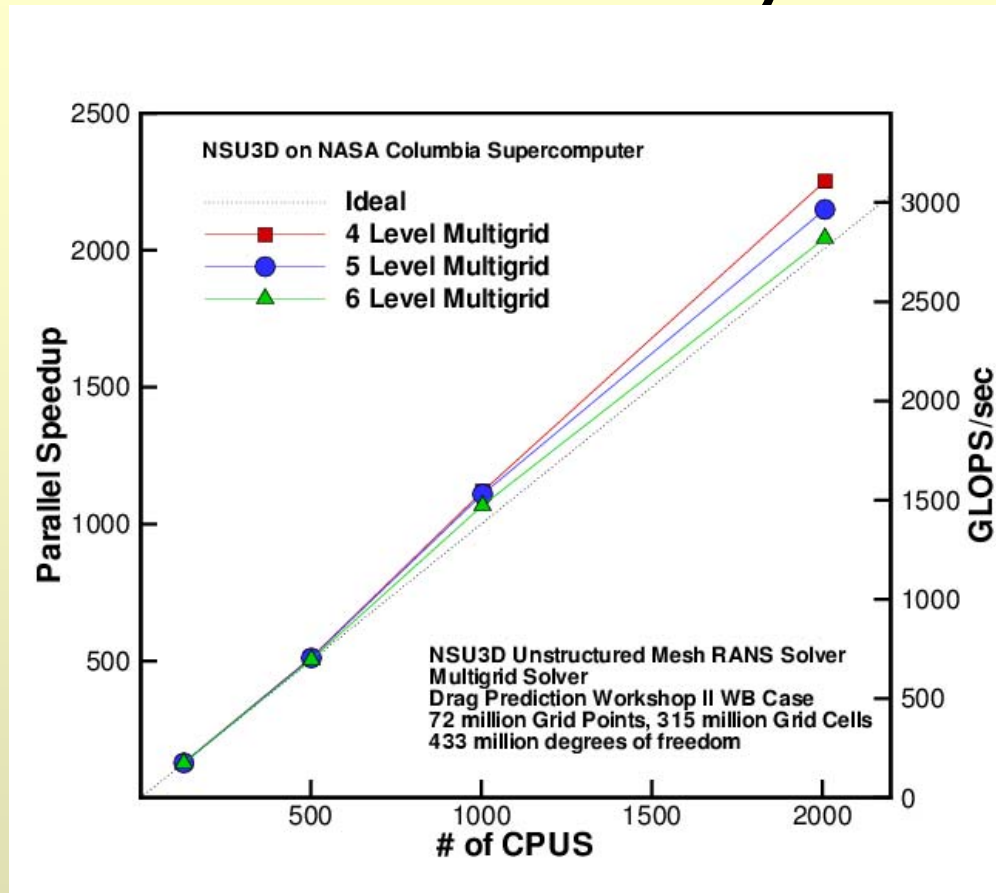
Partitioning Example

- 32-way partition of 30,562 point 2D grid



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Scalability



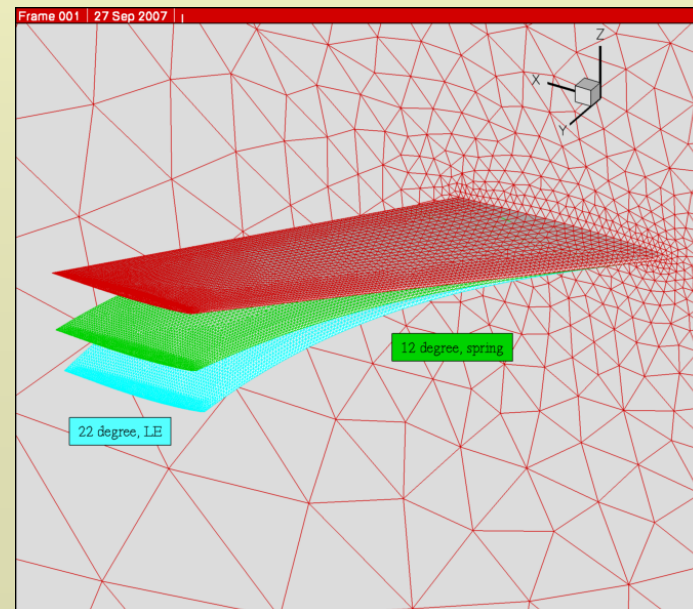
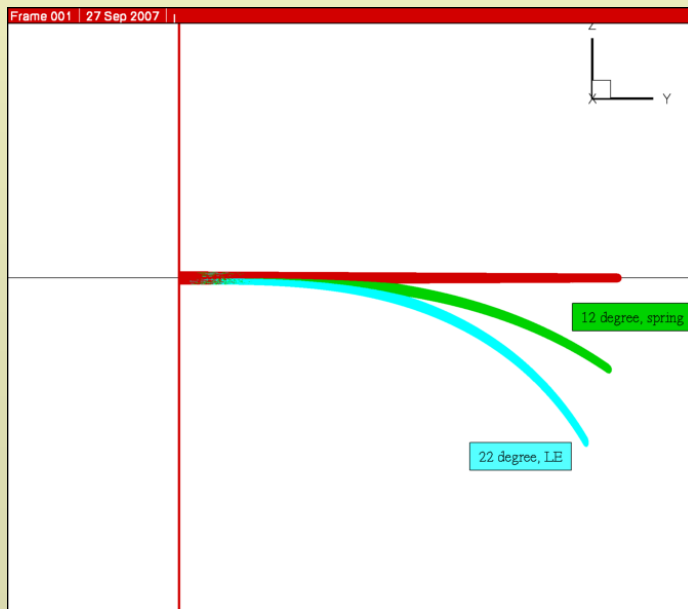
- Near ideal speedup for 72M pt grid on 2008 cpus of NASA Columbia Machine
 - Homogeneous Data-Structure
 - Near perfect load balancing
 - Near Optimal Partitioners

Mesh Motion

- Developed for MDO and Aeroelasticity Problems
- Emphasis on Robustness
 - Spring Analogy
 - Truss Analogy, Beam Analogy
 - Linear Elasticity: Variable Modulus
 - Regions of stiff E displace as solid body (no deformation)
- Emphasis on Efficiency
 - Edge based formulation
 - Gauss Seidel Line Solver with Agglomeration Multigrid

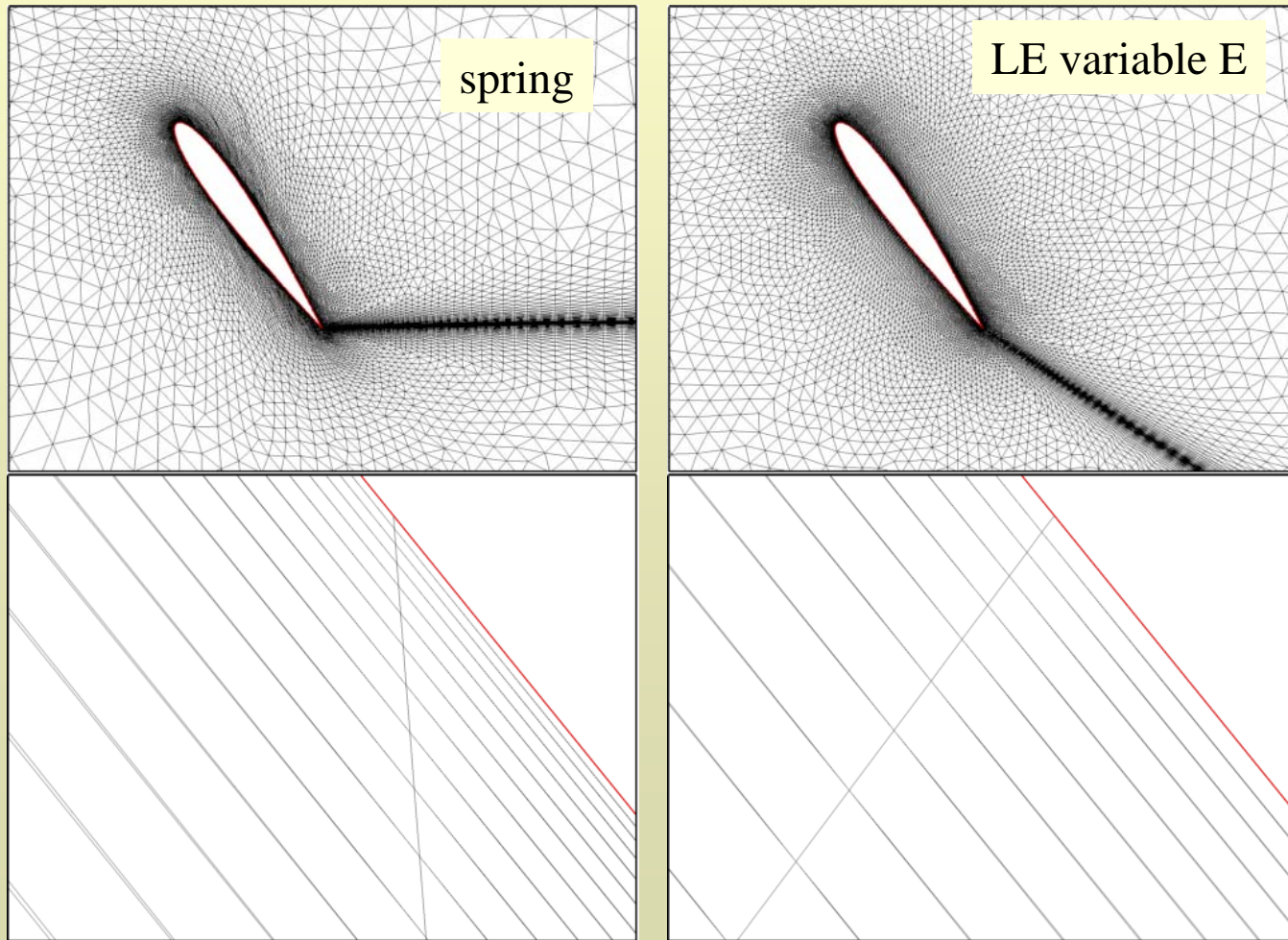
Mesh Deformation

- Mesh motion code tested for sample configurations
 - Bend until failure of mesh motion code
 - Spring analogy: Less robust, simpler
 - Linear elasticity analogy: More robust, more costly (x2)
 - Solved by multigrid in all cases



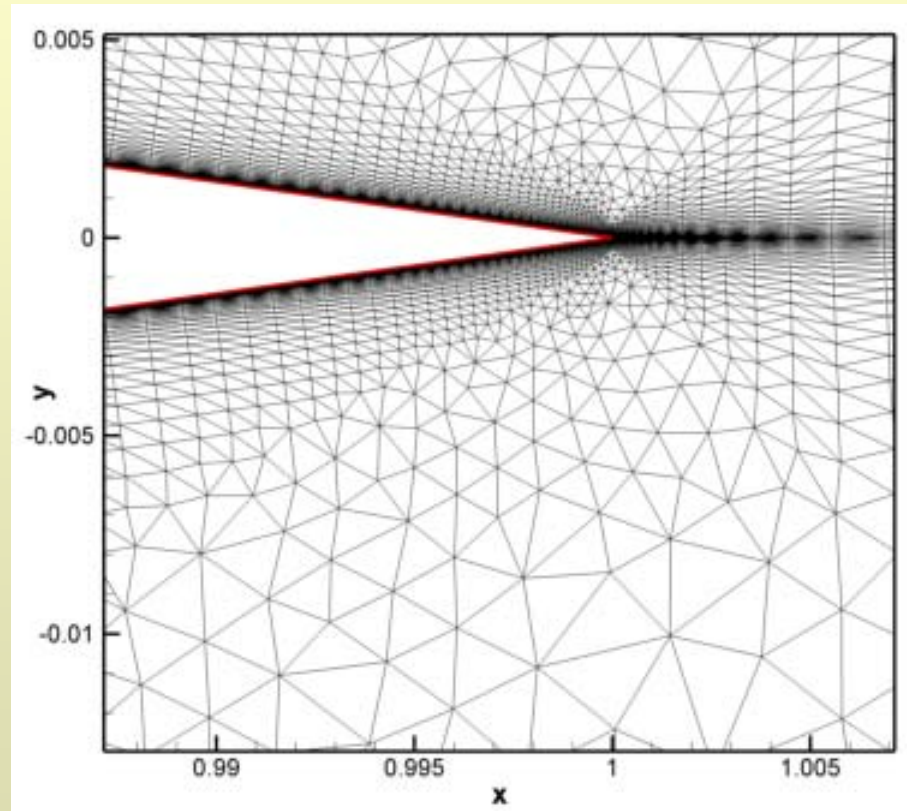
Mesh Motion Formulations

- *Linear elasticity approach preserves mesh orthogonality near surface (displaces mesh as solid body in regions of high E)*
- *Line solver displaces boundary layer mesh region rapidly in response to surface deflections*



Fast MG Solution of Mesh Motion Eqns

- Line solver + MG4 , first 10 iterations

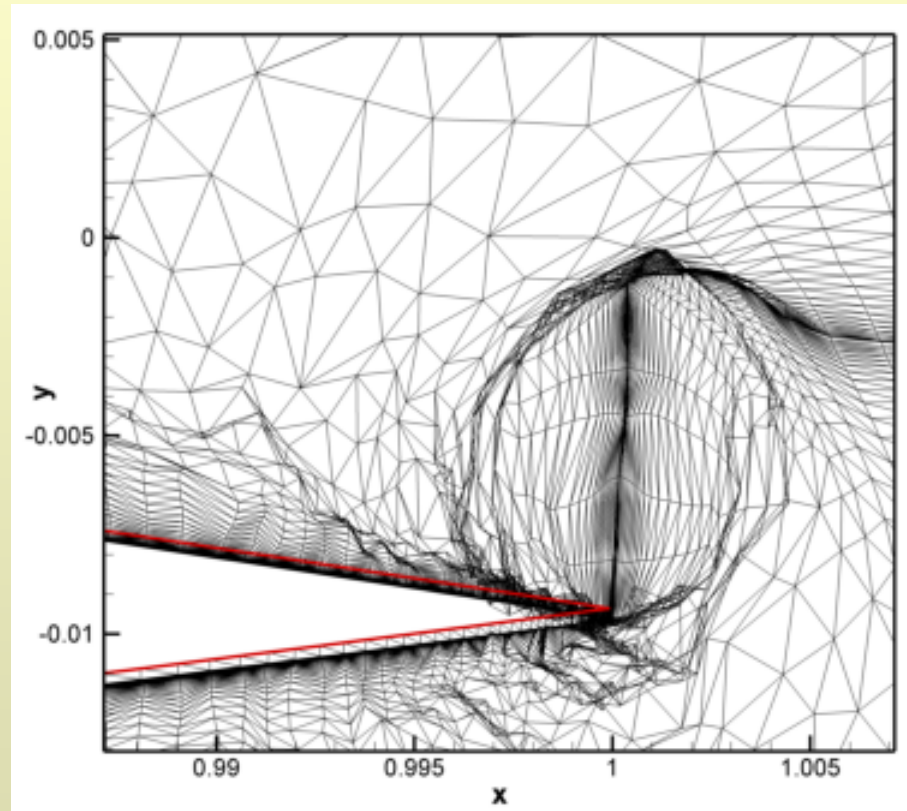


iter= 0

Viscous mesh, linear elasticity with variable E

Fast MG Solution of Mesh Motion Eqns

- Line solver + MG4 , first 10 iterations

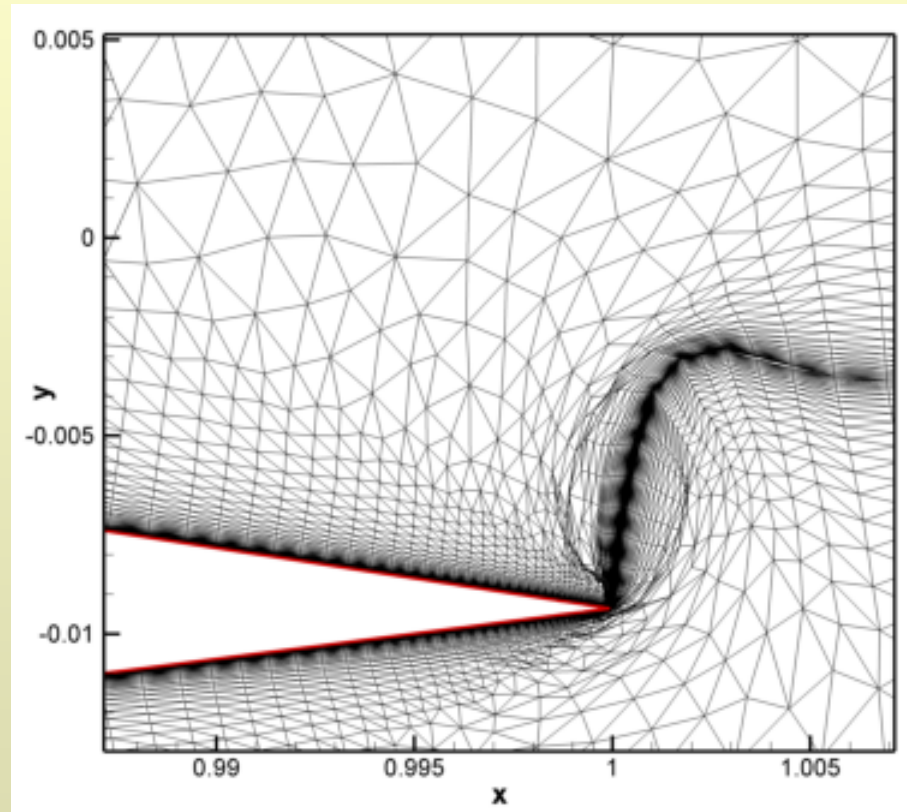


iter= 1

Viscous mesh, linear elasticity with variable E

Fast MG Solution of Mesh Motion Eqns

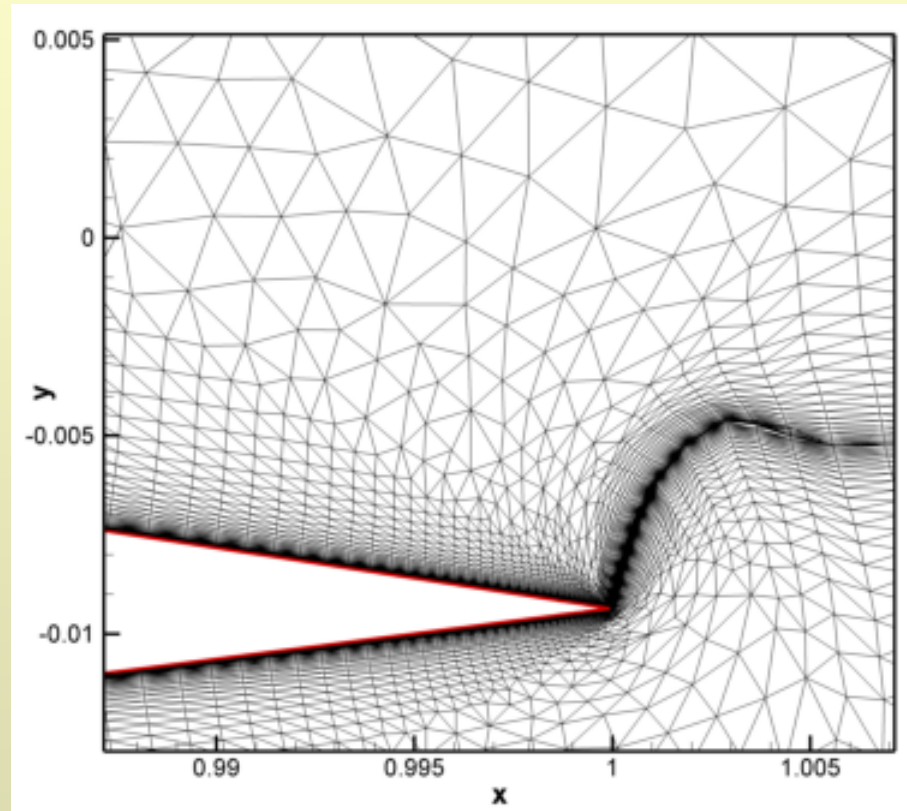
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Viscous mesh, linear elasticity with variable E

Fast MG Solution of Mesh Motion Eqns

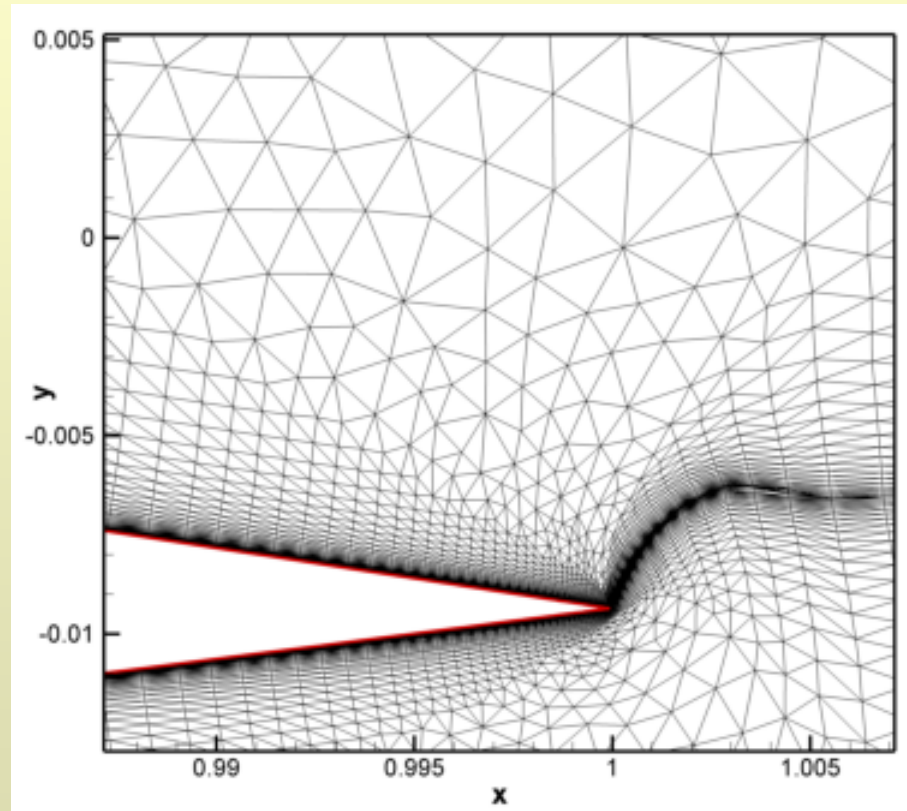
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Viscous mesh, linear elasticity with variable E

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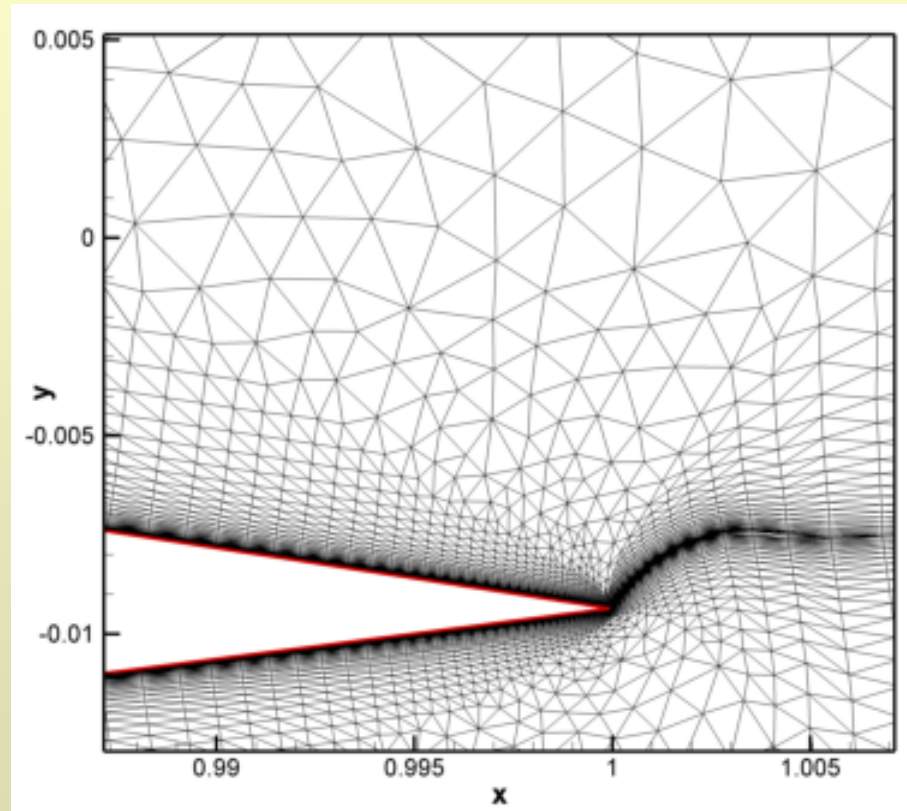


iter= 4

Viscous mesh, linear elasticity with variable E

Fast MG Solution of Mesh Motion Eqns

- Line solver + MG4 , first 10 iterations

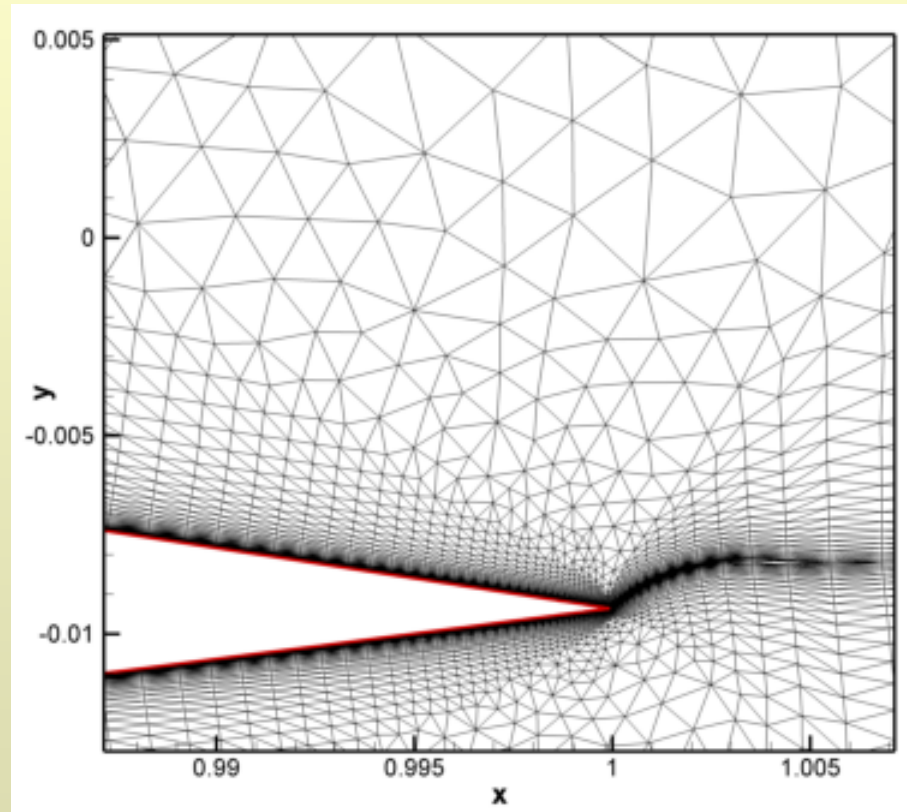


iter= 5

Viscous mesh, linear elasticity with variable E

Fast MG Solution of Mesh Motion Eqns

- Line solver + MG4 , first 10 iterations

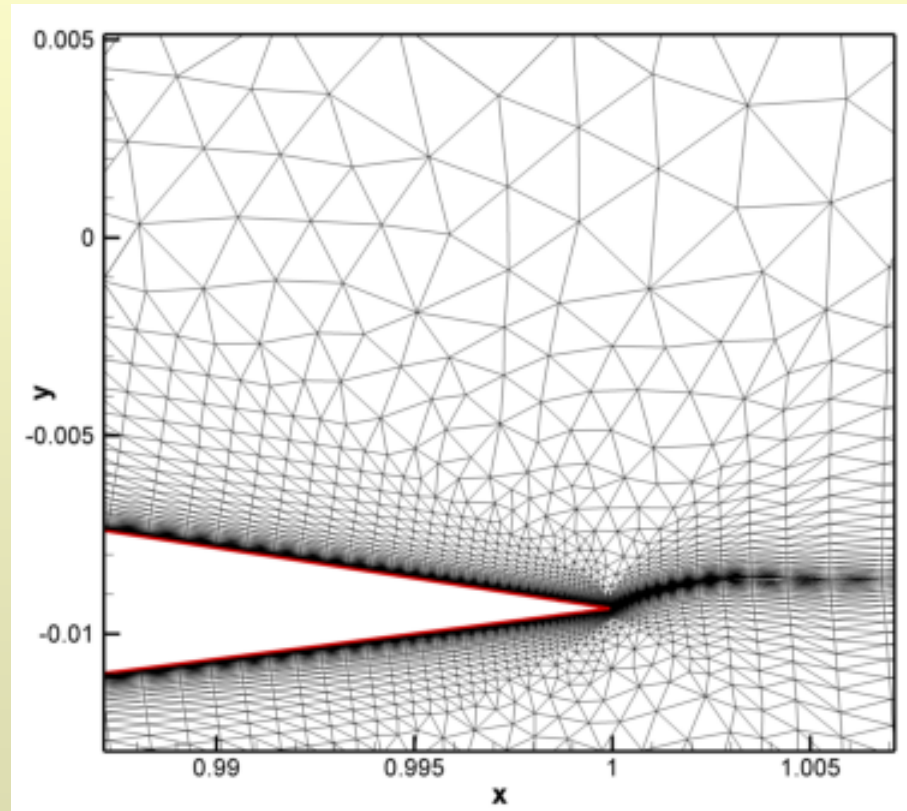


iter= 6

Viscous mesh, linear elasticity with variable E

Fast MG Solution of Mesh Motion Eqns

- Line solver + MG4 , first 10 iterations

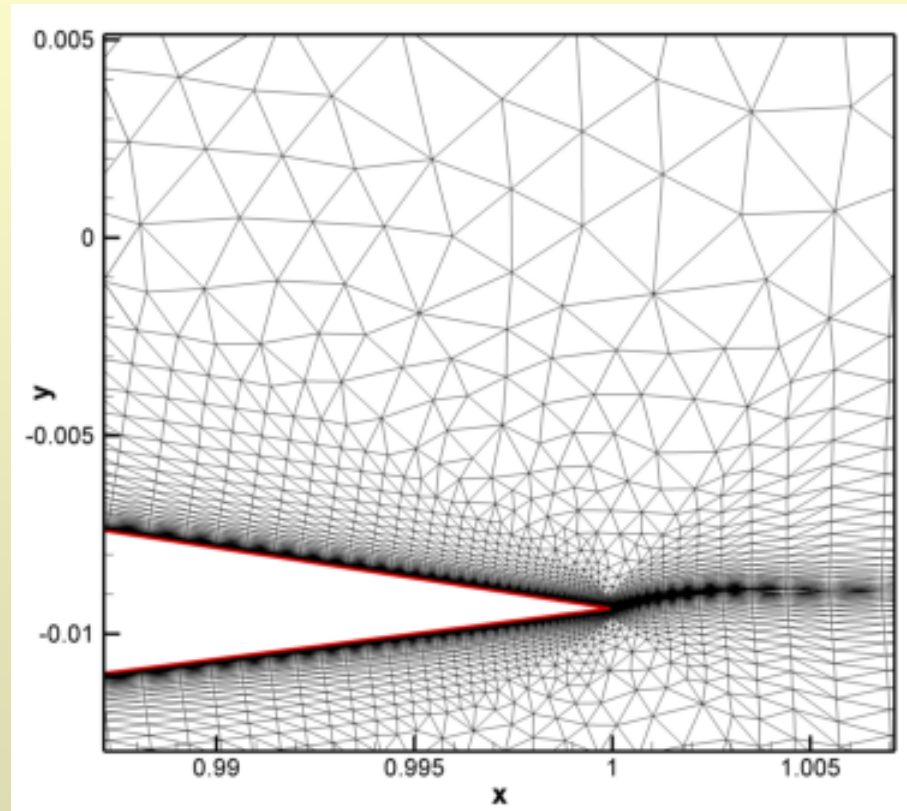


iter= 7

Viscous mesh, linear elasticity with variable E

Fast MG Solution of Mesh Motion Eqns

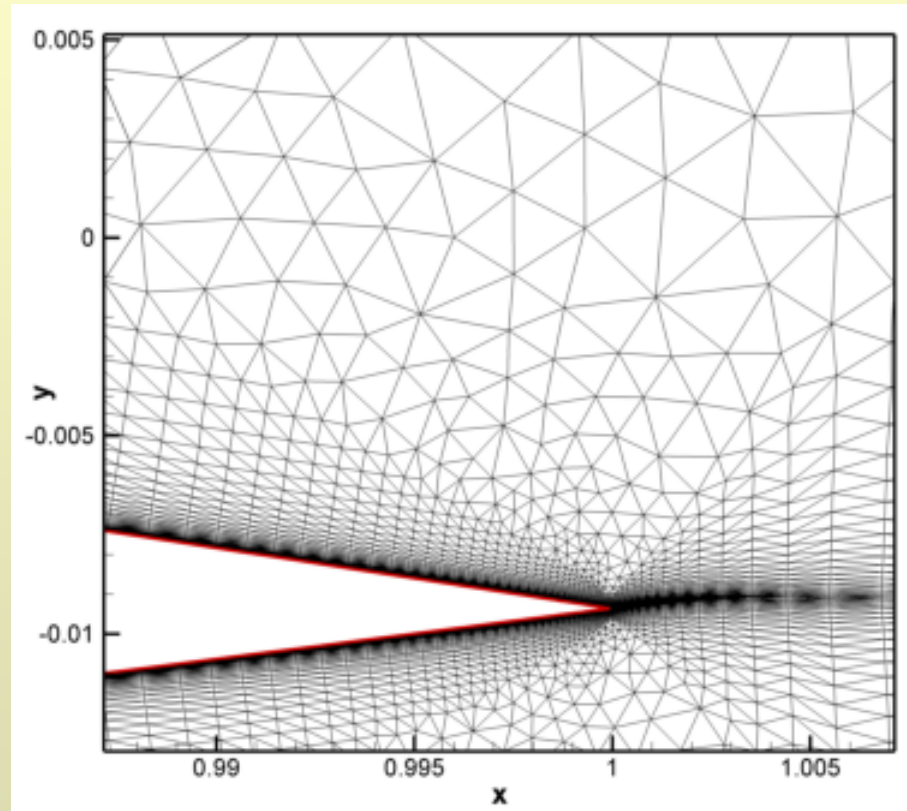
- Line solver + MG4 , first 10 iterations



Viscous mesh, linear elasticity with variable E

Fast MG Solution of Mesh Motion Eqns

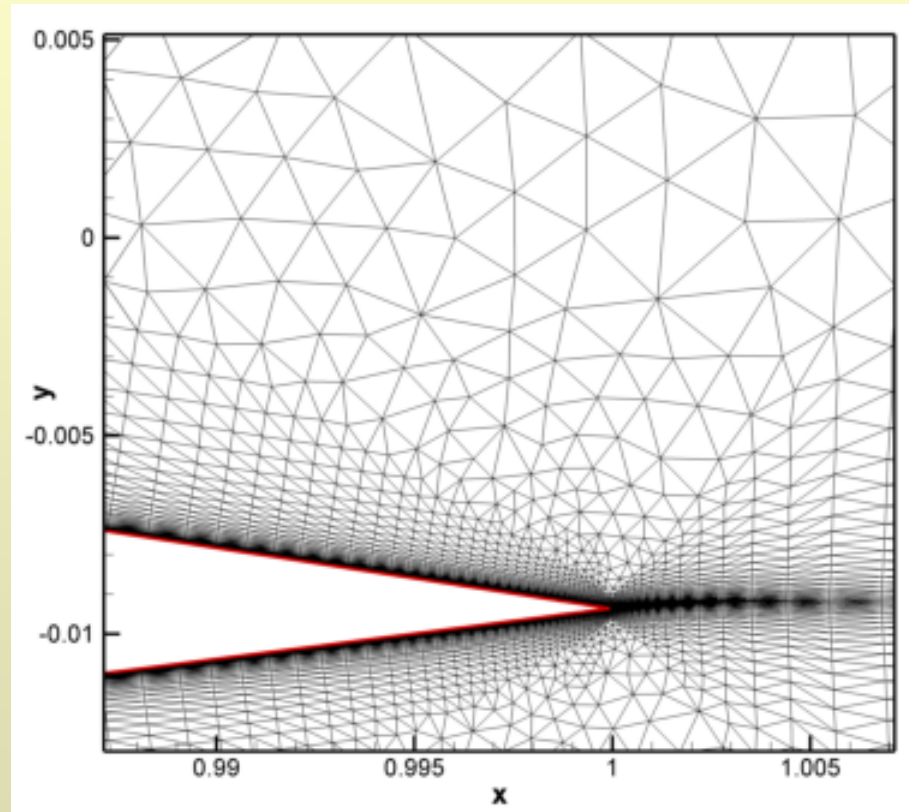
- Line solver + MG4 , first 10 iterations



Viscous mesh, linear elasticity with variable E

Fast MG Solution of Mesh Motion Eqns

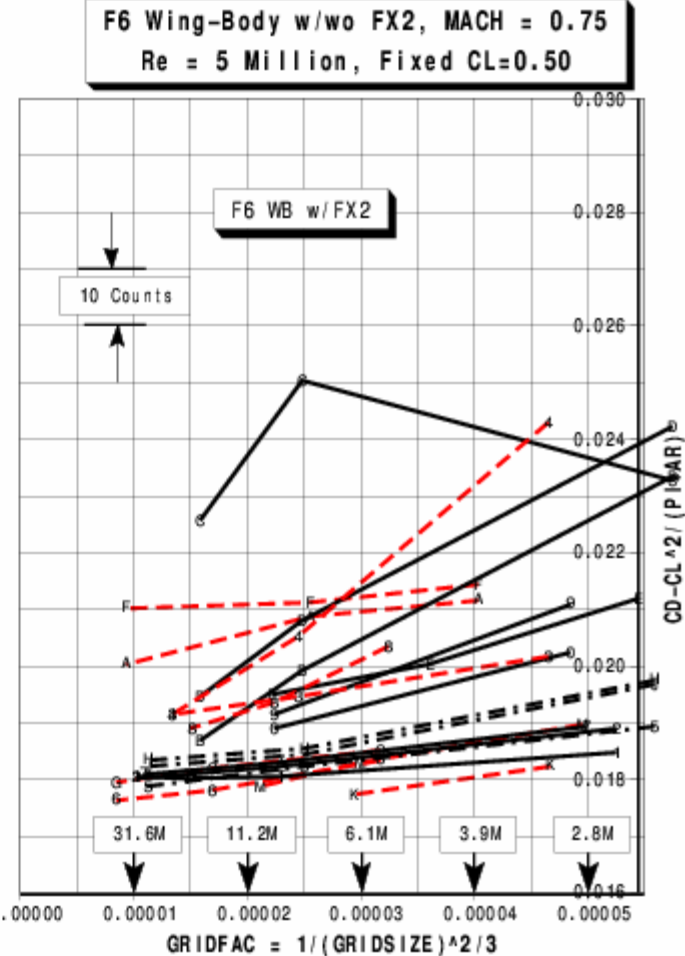
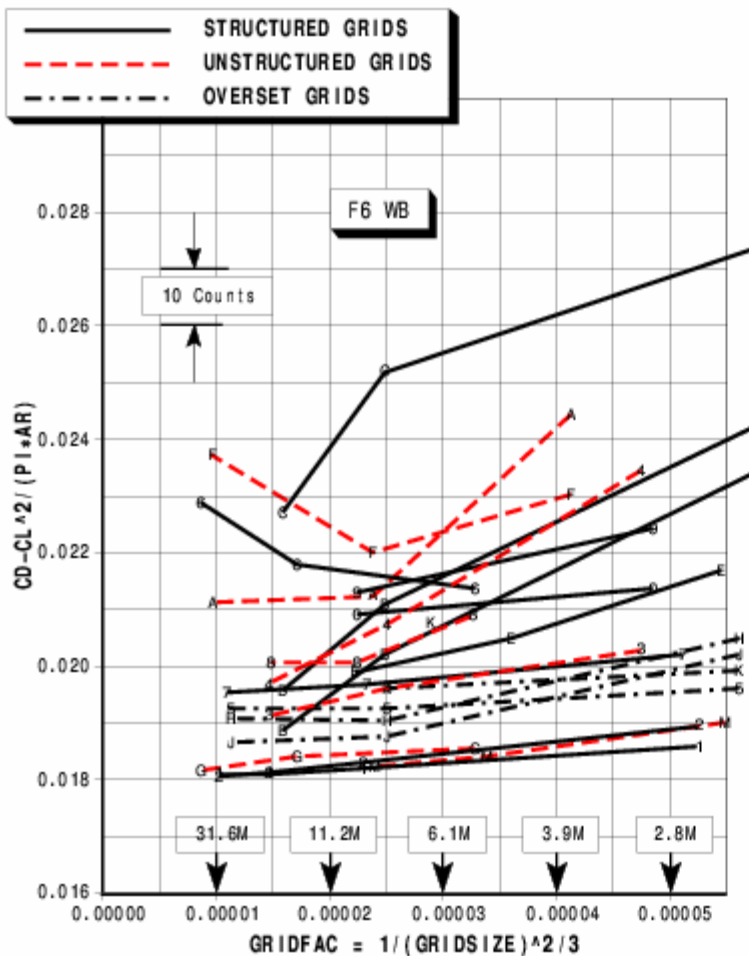
- Line solver + MG4 , first 10 iterations



Viscous mesh, linear elasticity with variable E

NSU3D=G
CFL3D=1,2
Overflow=H,K,...

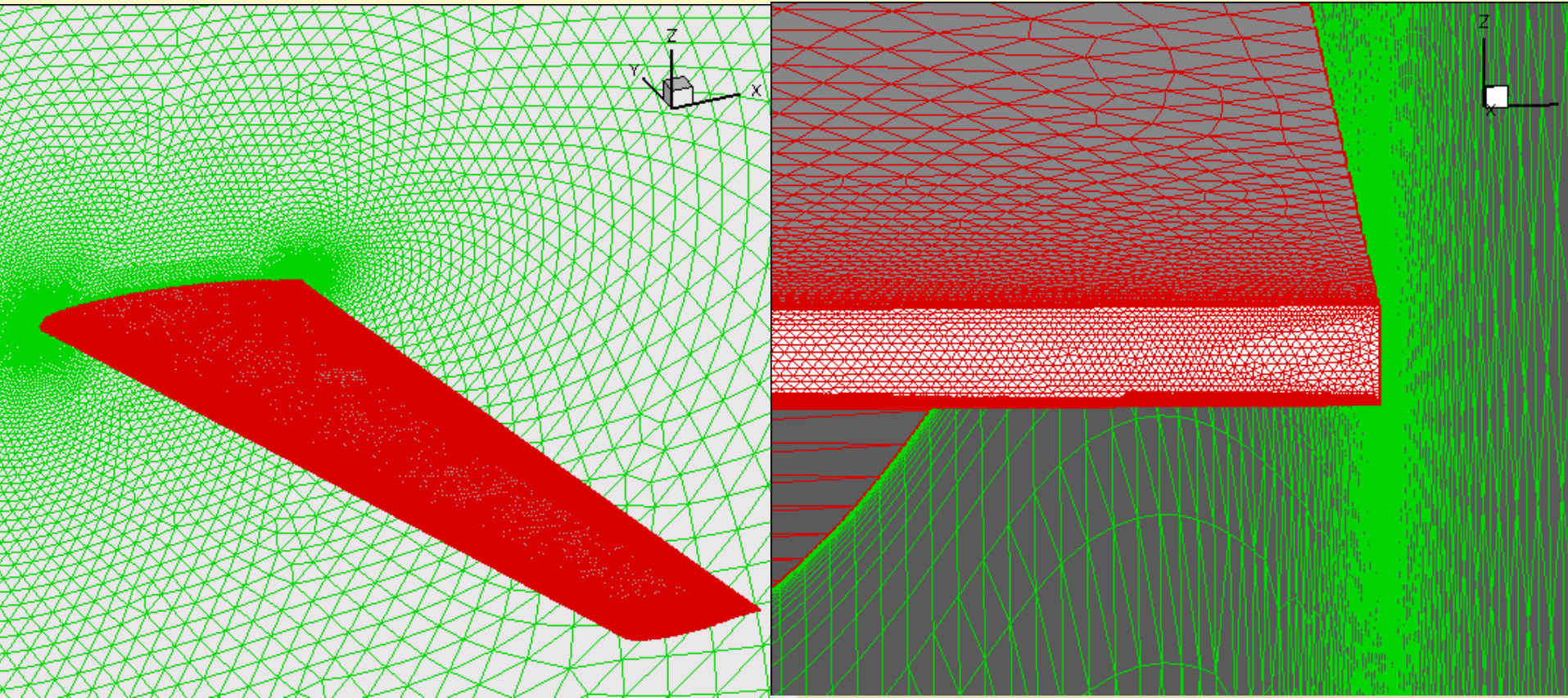
Grid Convergence – All Solutions



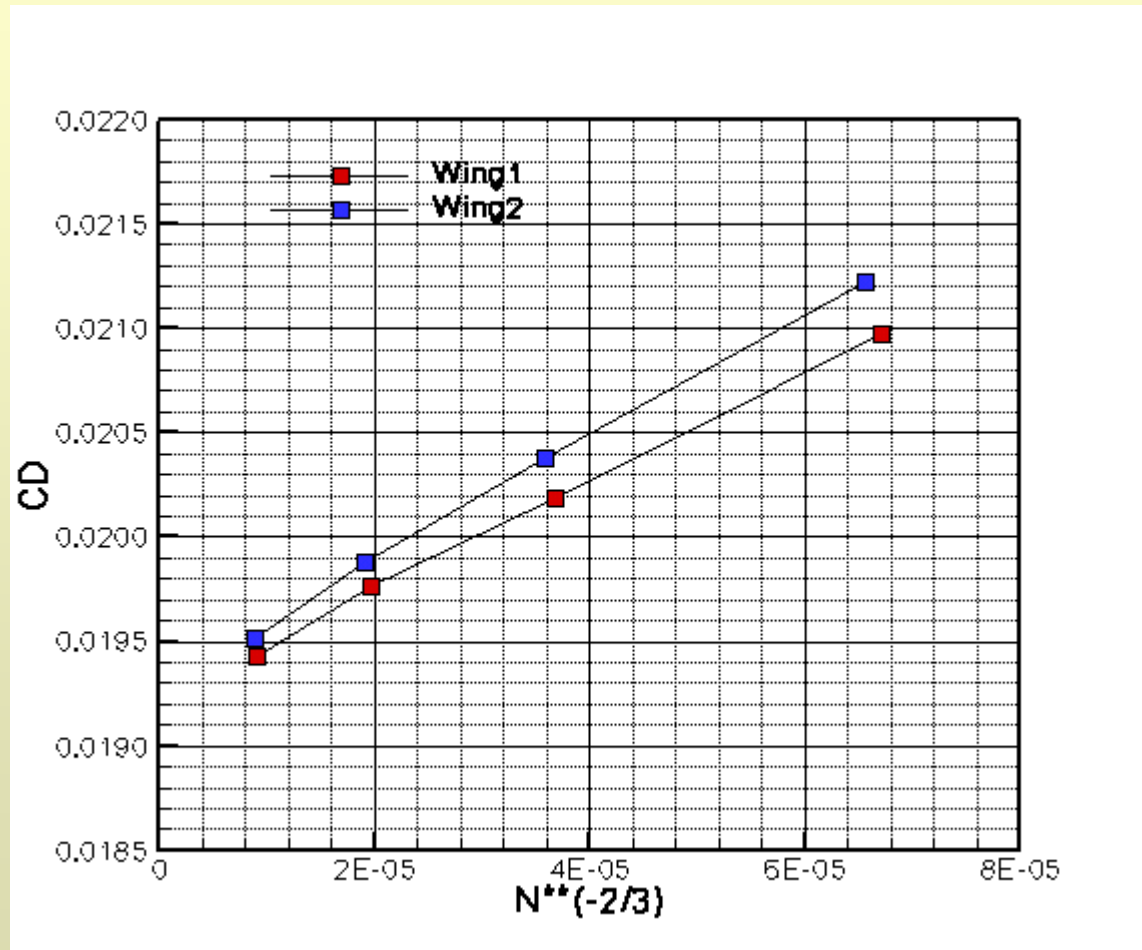
Unstructured vs Structured (Transonics)

- Considerable scatter in both cases
- No clear advantage of one method over the other in terms of accuracy
- DPW3 Observation:
 - Core set of codes which:
 - Agree remarkably well with each other
 - Span all types of grids
 - Structured, Overset, Unstructured
 - Have been developed and used extensively for transonic aerodynamics

DPW3 Wing1-Wing2 Cases

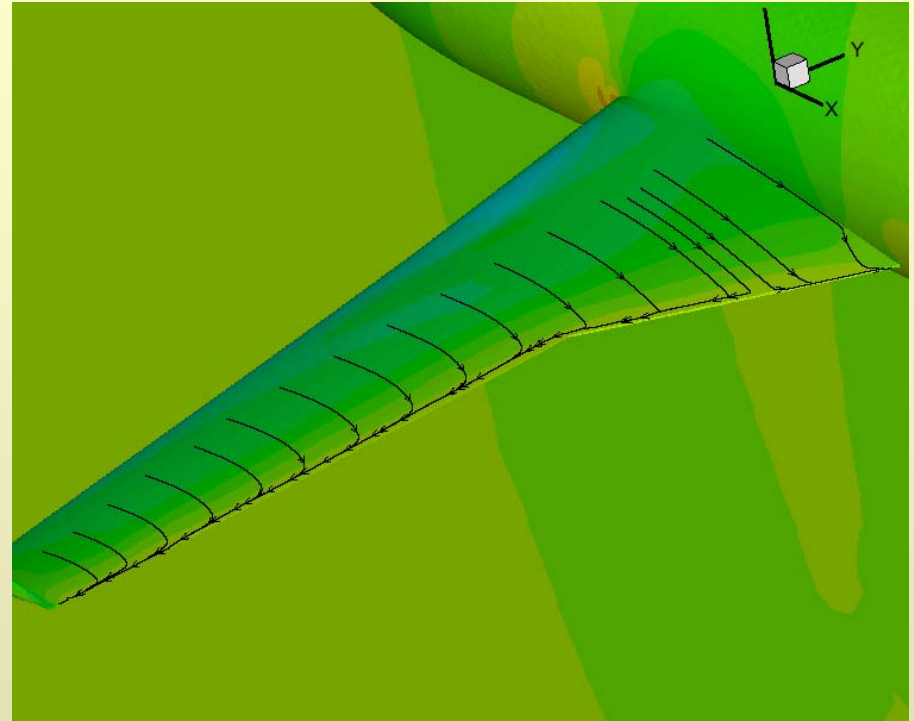
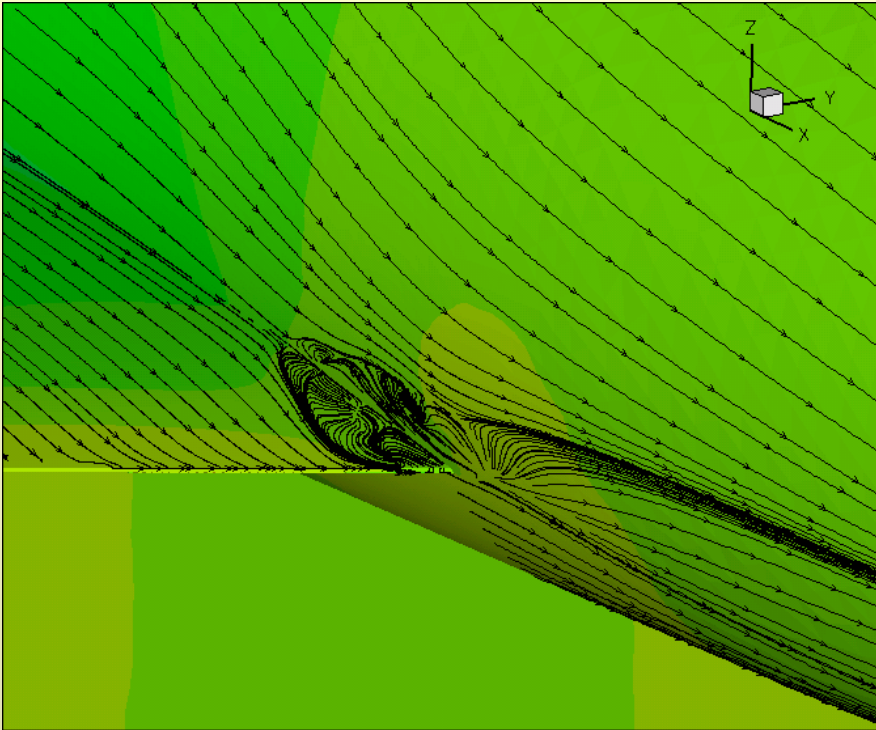


W1-W2 Grid Convergence Study (NSU3D)



- Apparently uniform grid convergence (attached flow cases)

Grid Resolution



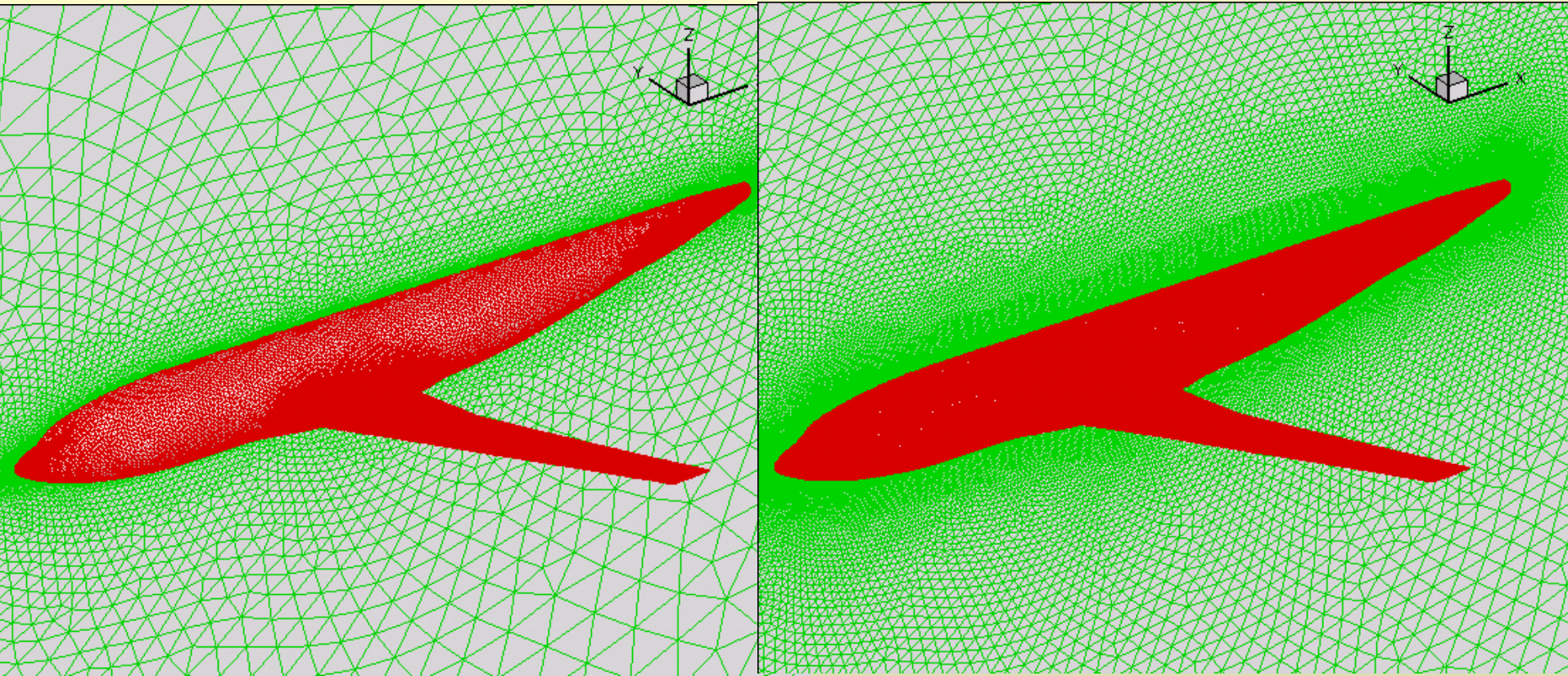
- Separated flow cases more demanding and often contradictory experiences

Grid Resolution Effects (DPW-3)

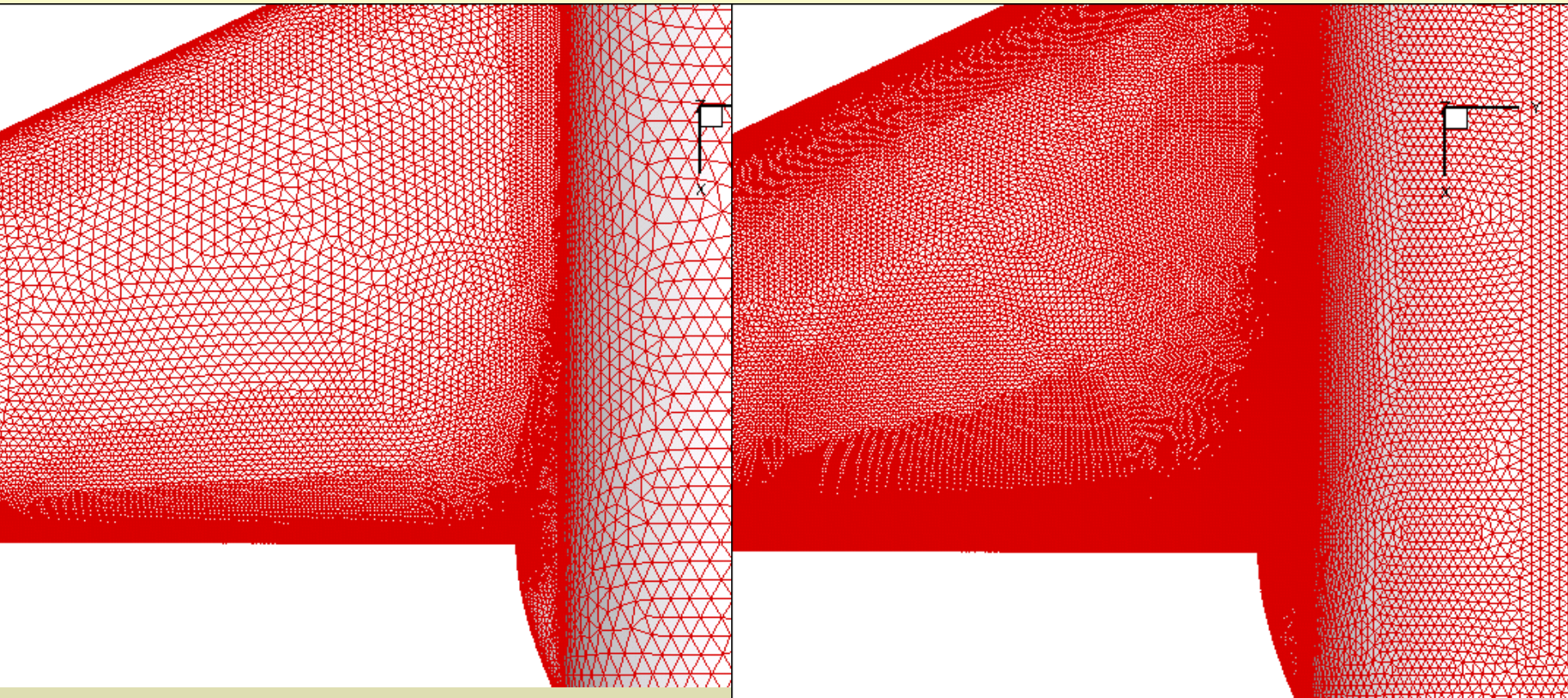
AIAA Paper 2008-0930

- SOB Separation increases with grid resolution
 - Boeing: Overset
 - Boeing: Unstructured
 - DLR: Unstructured
- SOB Separation remains constant with grid res.
 - Boeing: Block Structured
 - JAXA: Block Structured, Unstructured
- Trailing edge separation grows with grid res:
 - UW : Unstructured (NSU3D)
- Trailing edge separation constant with grid res:
 - JAXA: Structured, Unstructured
 - Boeing: Overset

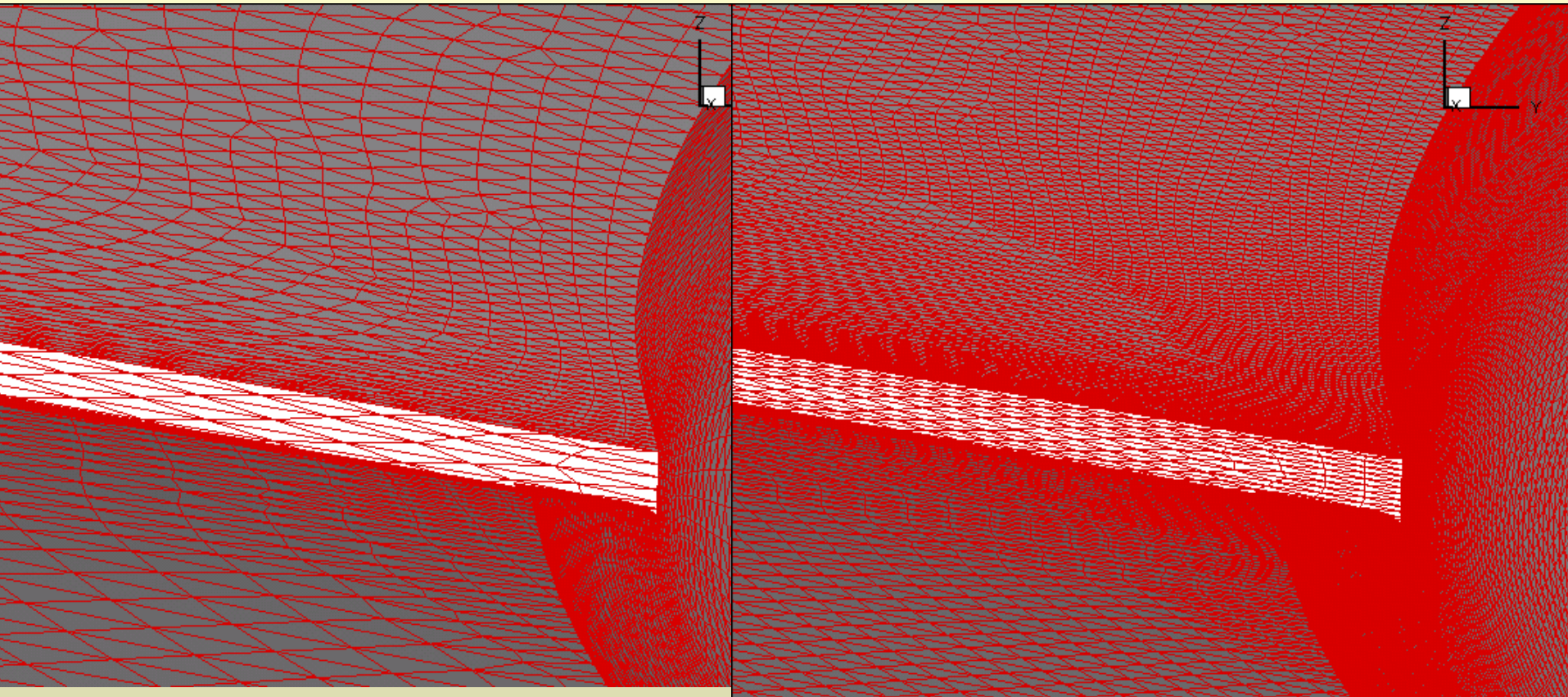
VGRID Node Centered (NASA)



VGRID Node Centered (NASA)



VGRID Node Centered (NASA)



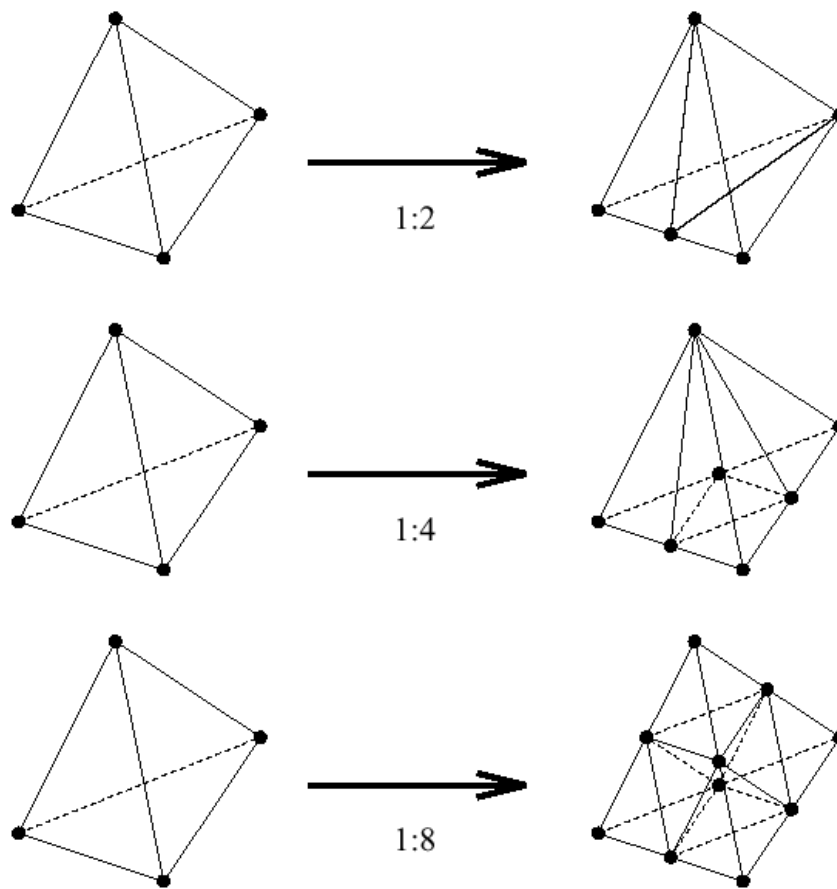
Grid Resolution

- Experimentation with much finer grids still required to understand behavior
 - Current largest grids ~ 100M pts
 - One more refinement ~ 10^9 pts
 - Implications for off-design conditions
 - Full flight envelope simulations
 - Stability and control
- AMR is obvious path for achieving higher resolution at manageable cost
 - Refinement criterion is roadblock
 - A-posteriori error estimation

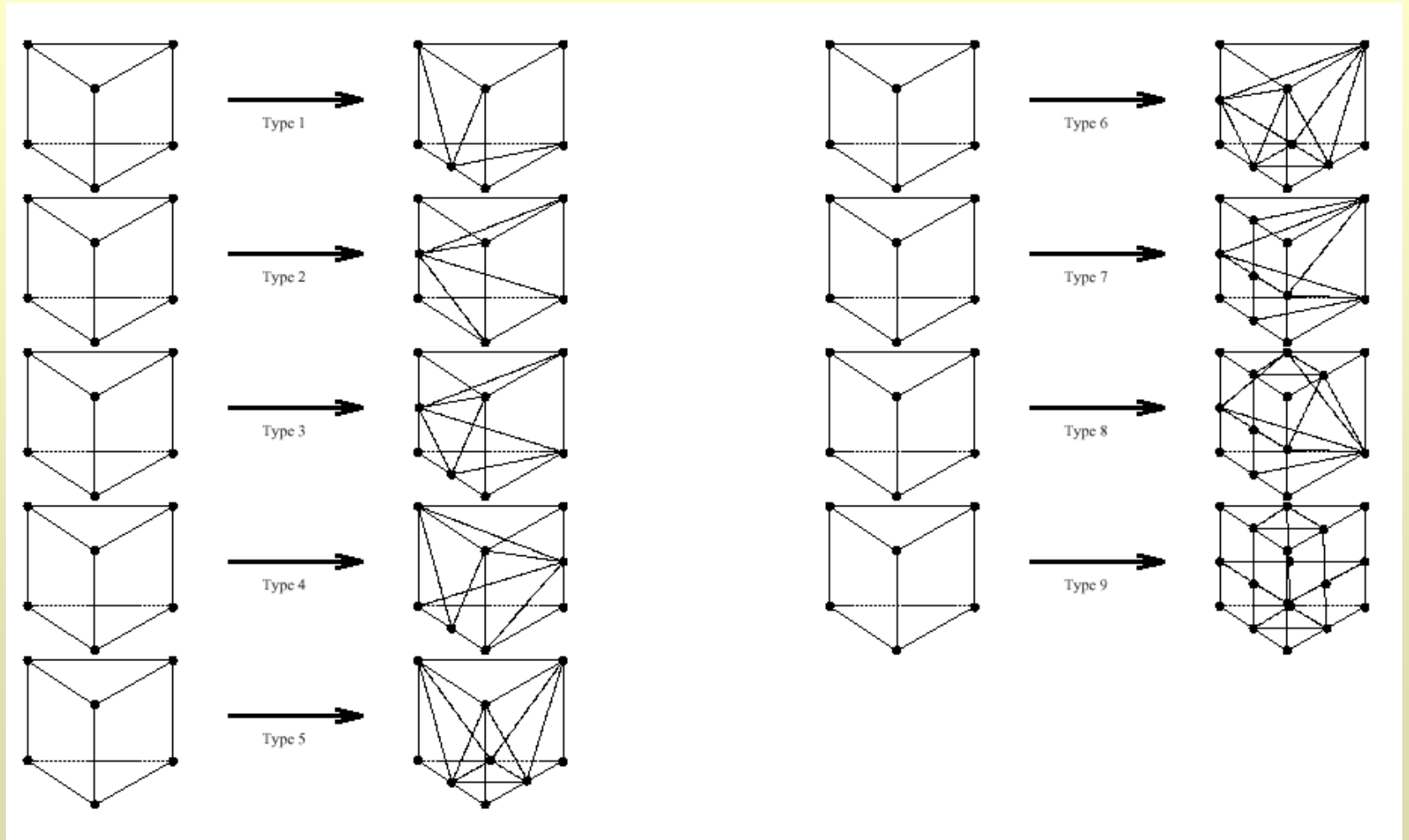
Adaptive Meshing Through Element Subdivision

- Standalone (sequential) code
- Flag Cells with Large Errors
- Subdivide these Cells (1:8)
- Adjust Cells in Transition Regions between Refined and Non-Refined Regions
 - (anisotropically refined cells)

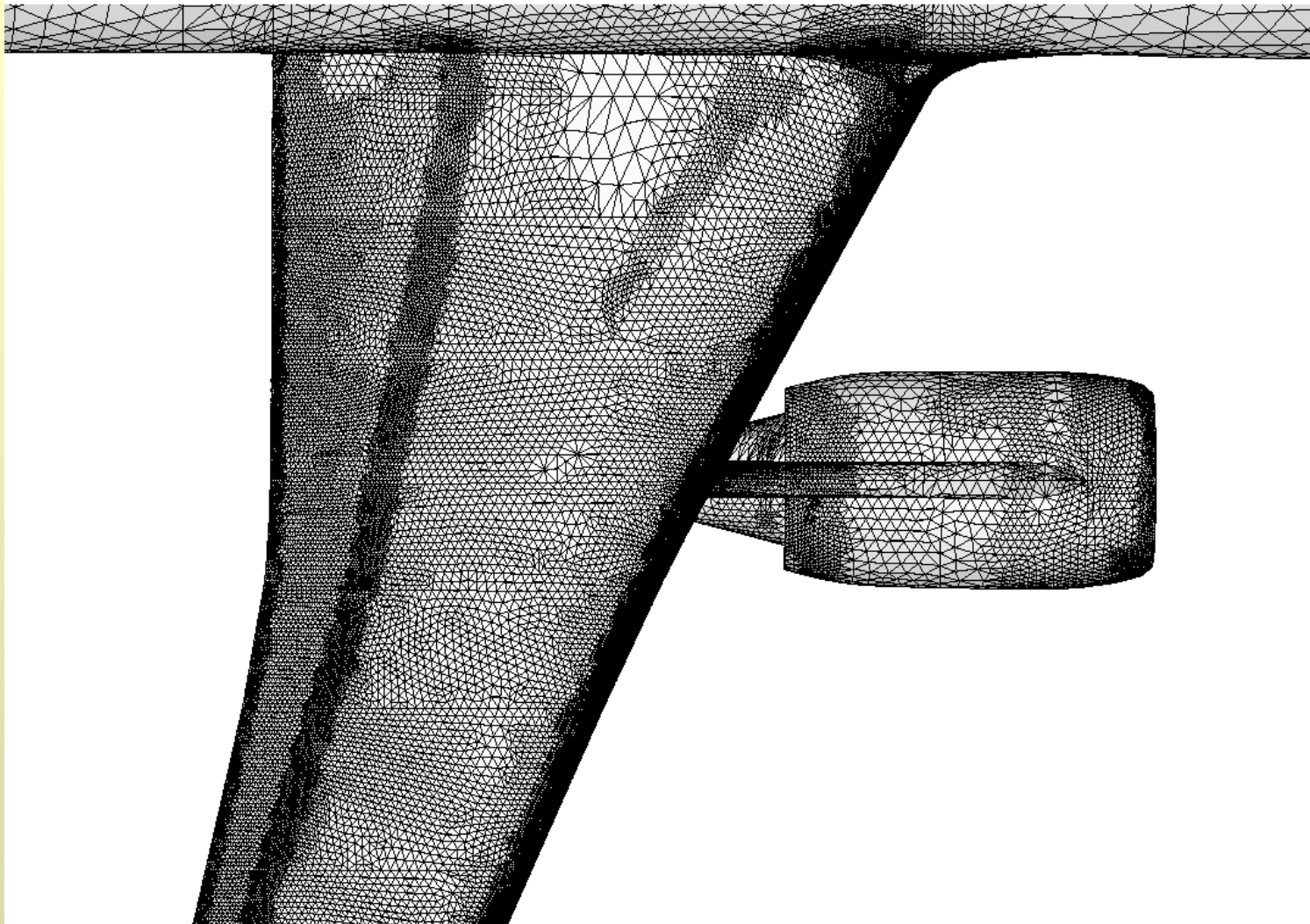
Subdivision Types for Tetrahedra



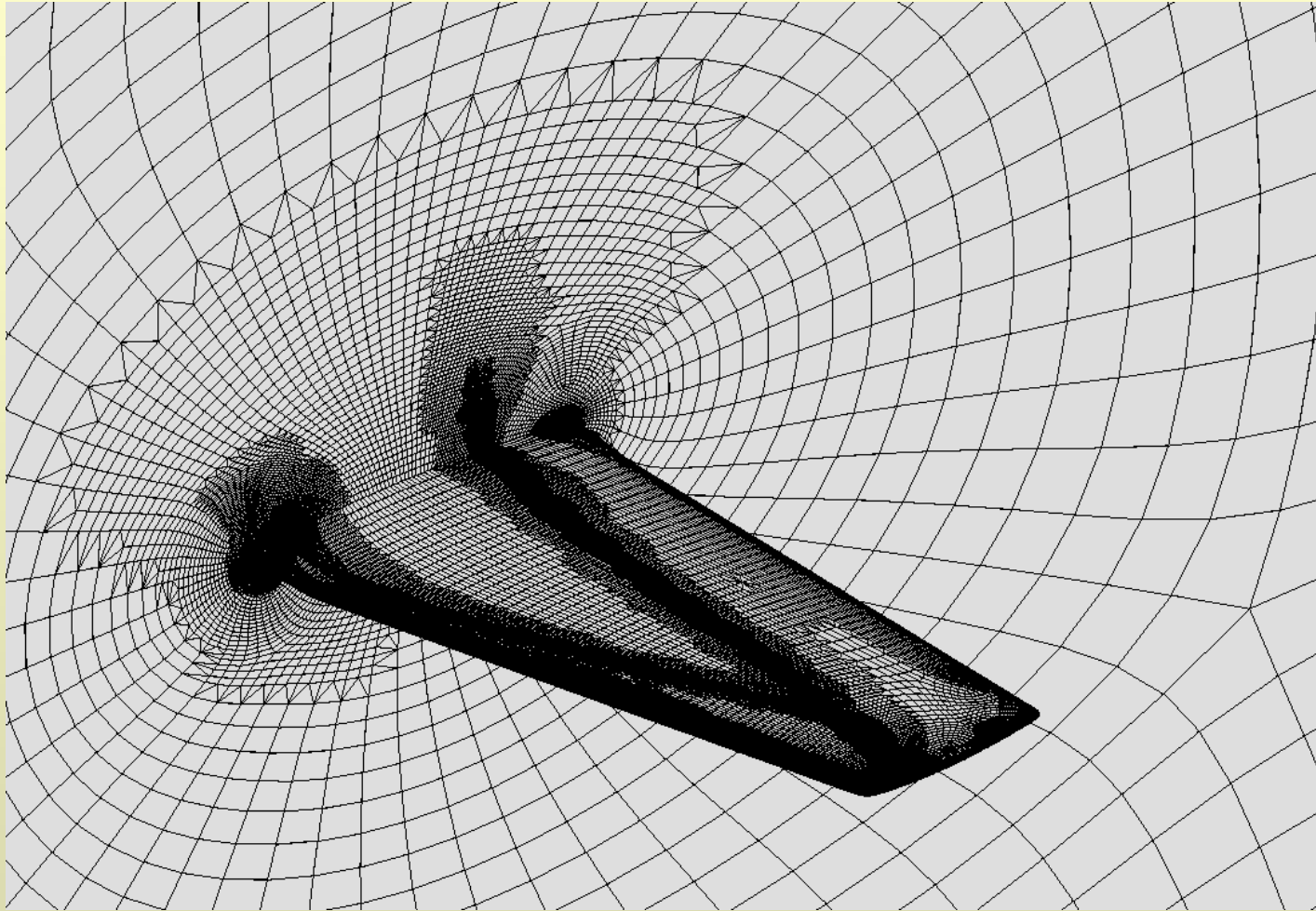
Subdivision Types for Prisms



Adaptive Tetrahedral Mesh by Subdivision



Adaptive Hexahedral Mesh by Subdivision



Adjoint-Based Error Estimation

- Complex simulations have multiple error sources
- Engineering simulations concerned with specific output objectives
- Adjoint methods / Goal Oriented Approach
 - Methodical approach for constructing discrete adjoint
 - Use for a posteriori error estimation
 - Spatial error
 - Temporal error
 - Other error sources
 - Use to drive adaptive process

Spatial Error Sensitivity

Fine level Taylor expansion of functional objective L:

$$L_h(U_h) = L_h(U_h^H) + \left[\frac{\partial L}{\partial U} \right] (U_h - U_h^H)$$

h = fine grid (solution not available)

H = coarse grid (solution available)

Spatial Error Sensitivity

Fine level flow residual Taylor expansion

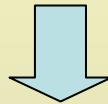
$$R_h(U_h) = R_h(U_h^H) + \left[\frac{\partial R}{\partial U} \right]_{U_h^H} (U_h - U_h^H) = 0$$

$$U_h - U_h^H = - \left[\frac{\partial R}{\partial U} \right]^{-1} R_h(U_h^H)$$

Spatial Error Sensitivity

Fine level Taylor expansion of functional objective L:

$$L_h(U_h) = L_h(U_h^H) - \left[\frac{\partial L}{\partial U} \right] \left[\frac{\partial R}{\partial U} \right]^{-1} R_h(U_h^H)$$



$$L_h(U_h) - L_h(U_h^H) = -\Lambda^T \cdot R_h(U_h^H)$$

$$\left[\frac{\partial R}{\partial U} \right]^T \Lambda = - \left[\frac{\partial L}{\partial U} \right]^T \quad \text{Adjoint problem}$$

h = fine grid (solution not available)

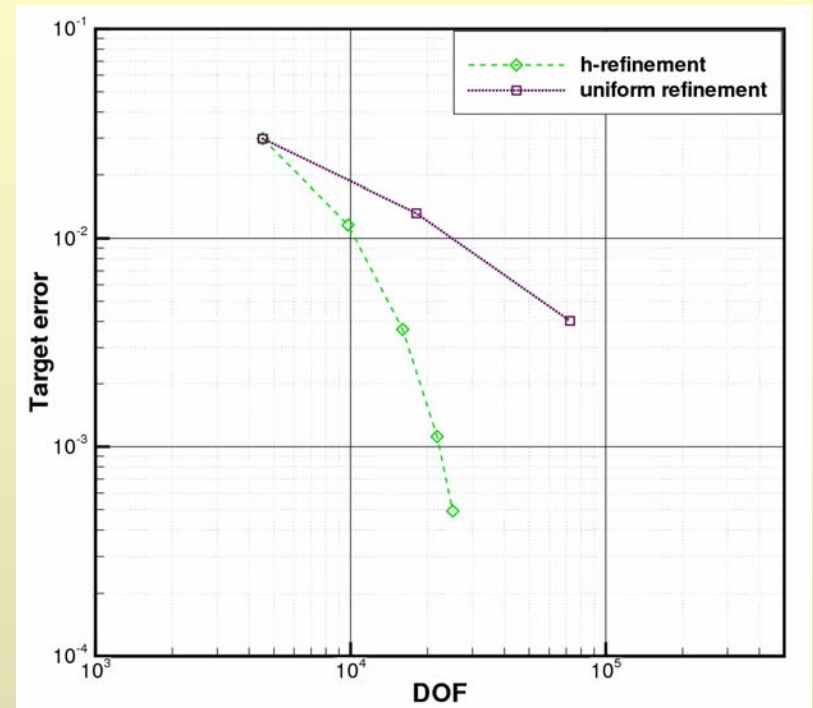
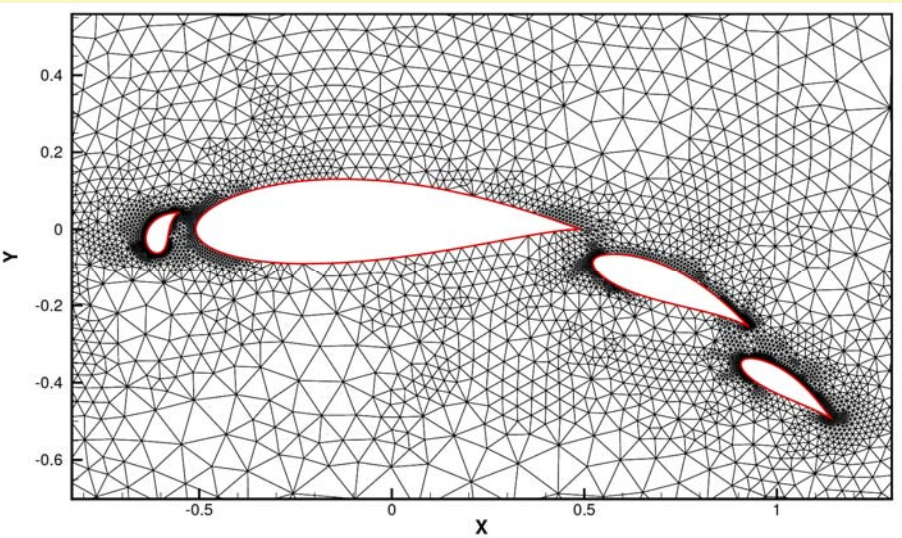
H = coarse grid (solution available)

Spatial Error Sensitivity

$$L_h(U_h) - L_h(U_h^H) = -\Lambda^T \cdot R_h(U_h^H)$$

- Compute objective on current grid
- Compute adjoint on current grid
- Project solution on to finer grid, evaluate residual (non-zero)
- Change in Objective on fine grid is inner product of adjoint with residual on fine grid
 - Cost: 1 flow solve, 1 adjoint solve on coarse grid
 - Gain: Prediction of objective on fine grid (8 times more grid points)
 - Use to certify sensitivity of solution to finer grid
 - Use to drive adaptive meshing (for specific objective)

Goal-Oriented Spatial Adaptivity (steady-state)



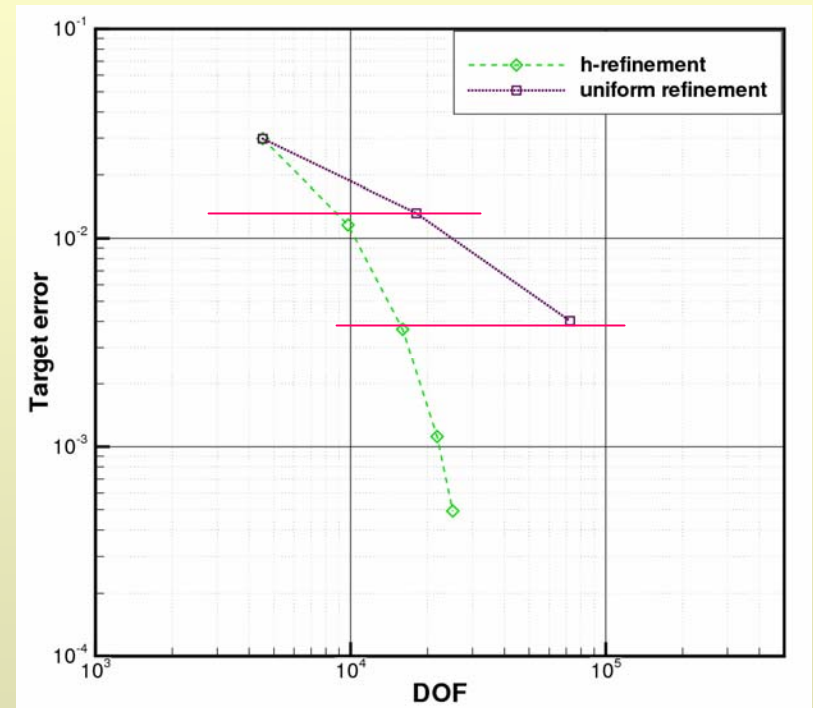
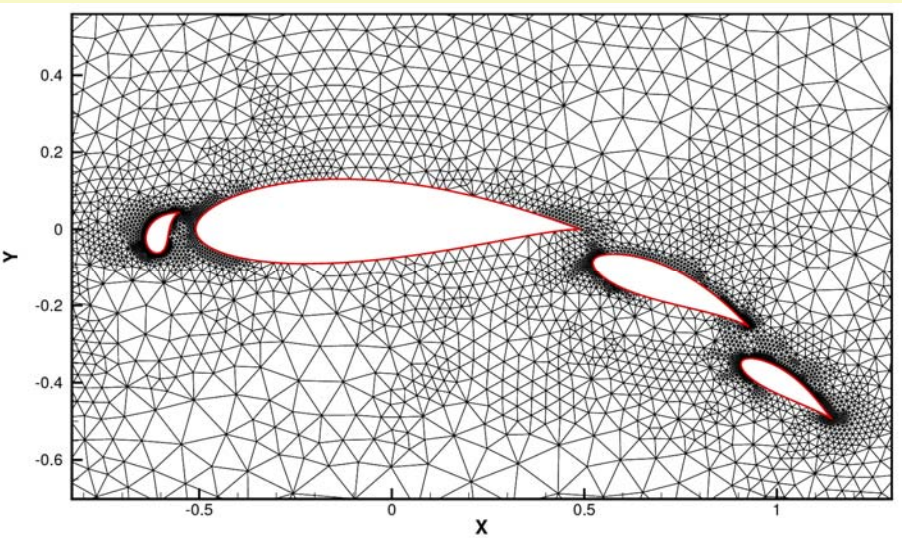
Adaptivity based on Drag

See also: Venditti and Darmofal; AIAA-2001

Giles et al; AIAA-2001

Park, M. AIAA-2002

Goal-Oriented Spatial Adaptivity (steady-state)



Adaptivity based on Drag

See also: Venditti and Darmofal; AIAA-2001

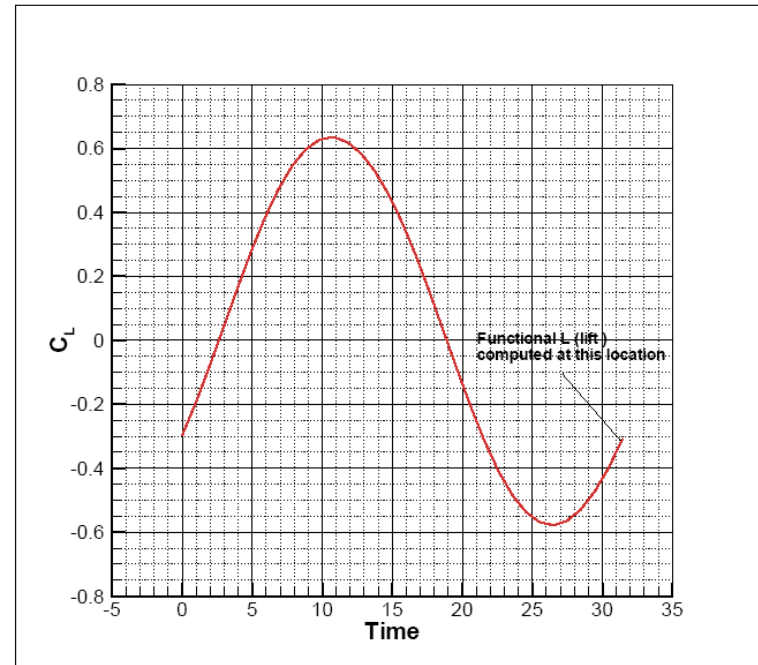
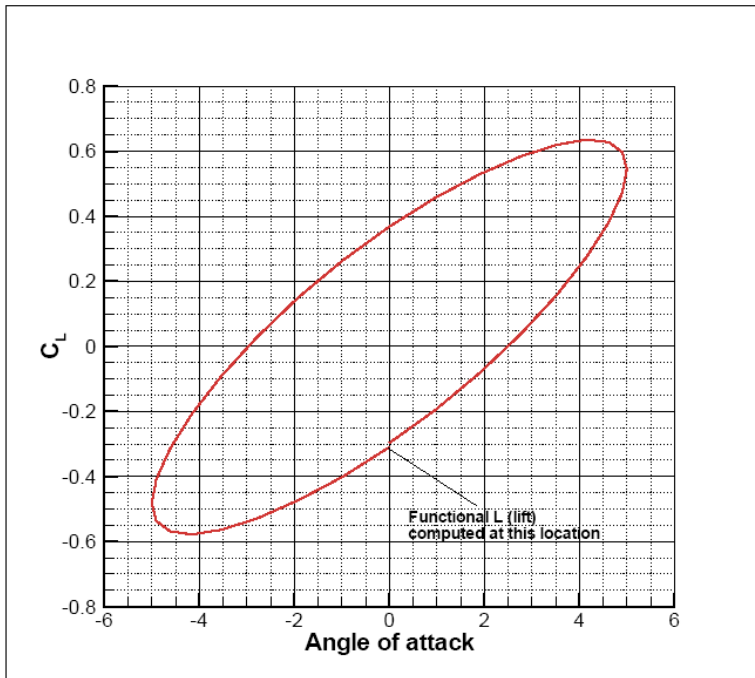
Giles et al; AIAA-2001

Park, M. AIAA-2002

Error Estimation for Time Dependent Problems

Test Case Description

Sinusoidally pitching airfoil
Functional scalar is Lift after 1 period



Easily extended to estimate error in time-integrated Lift history

Summary of Temporal Resolution Error Evaluation

- Compute unsteady flow solution on coarse time domain
- Compute adjoint variables on coarse time domain
 - Integrating backward in time
- Project adjoint variables, flow solution and mesh solution onto fine time domain
- Temporal resolution error is then inner product of adjoint with corresponding non-zero residual on fine time domain
 - Distribution in time is used to drive adaptation

Validation

Description	Functional value	% Error vs. Target	% Predicted Error
Target functional - exact at 32 steps (fully converged)	-0.309957065250867	-	-
Fully converged flow and mesh at 16 steps	-0.285768366164898	+7.804	+7.463
Corrected for resolution from 16 to 32 steps	-0.3089006774509025	+0.341	-

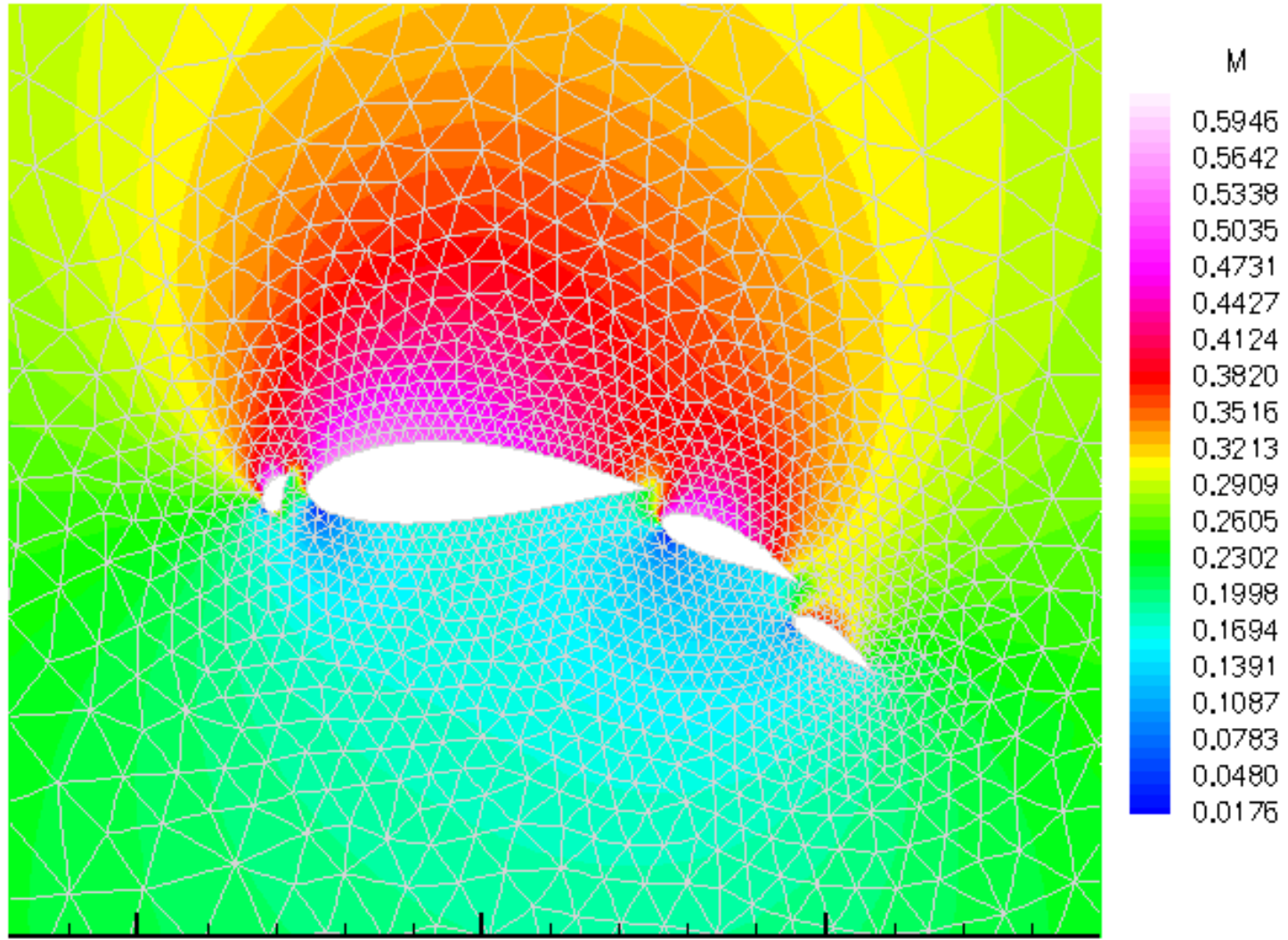
- Adjoint is linearization about current state (16 time steps) to predict objective value on modified state (32 time steps)

High Order Methods

- Higher order methods such as Discontinuous Galerkin best suited to meet high accuracy requirements
 - Asymptotic properties
- HOMs reduce grid generation requirements
- HOMs reduce grid handling infrastructure
 - Dynamic load balancing
- Compact data representation (data compression)
 - Smaller number of modal coefficients versus large number of point-wise values
- HOMs scale very well on massively parallel architectures

4-Element Airfoil (Euler Solution)

$P = 3$



Single Grid Steady-State Implicit Solver

- Steady state

$$\boxed{\mathbf{R}_p(\mathbf{U}_p) = \mathbf{S}_p}$$

- Newton iteration

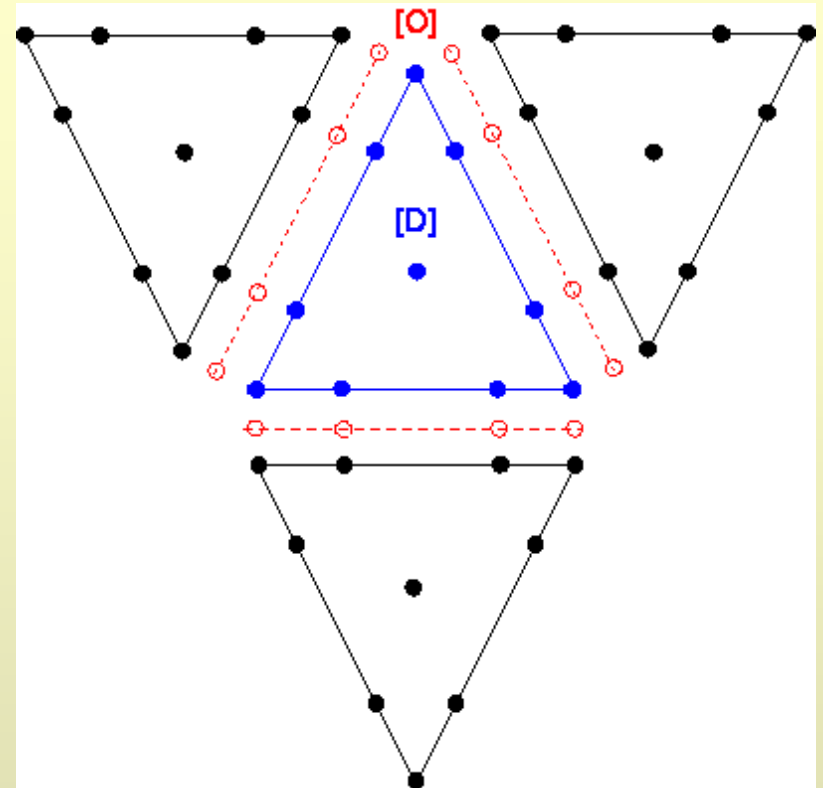
$$\left[\frac{\partial \mathbf{R}_p}{\partial \mathbf{U}_p} \right]^n \Delta \mathbf{U}_p^{n+1} = \mathbf{S}_p - \mathbf{R}_p(\mathbf{U}_p^n)$$

- Non-linear update

$$\mathbf{U}_p^{n+1} = \mathbf{U}_p^n + \Delta \mathbf{U}_p^{n+1}$$

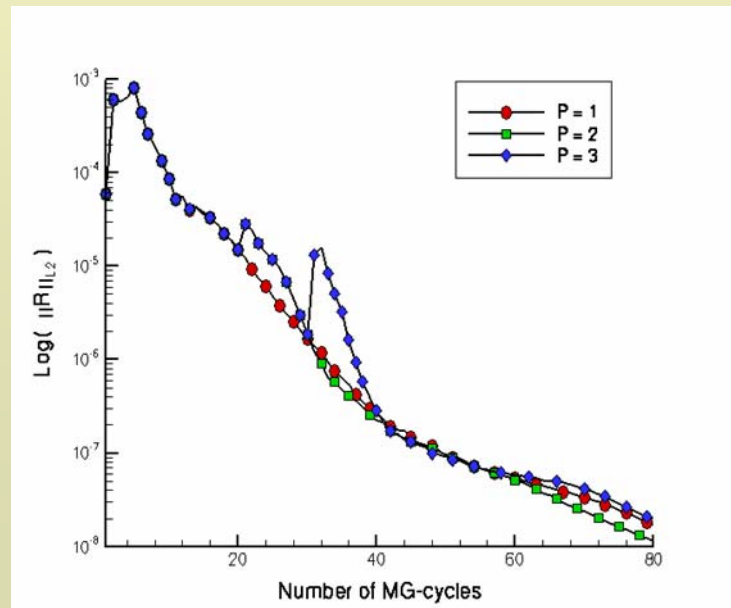
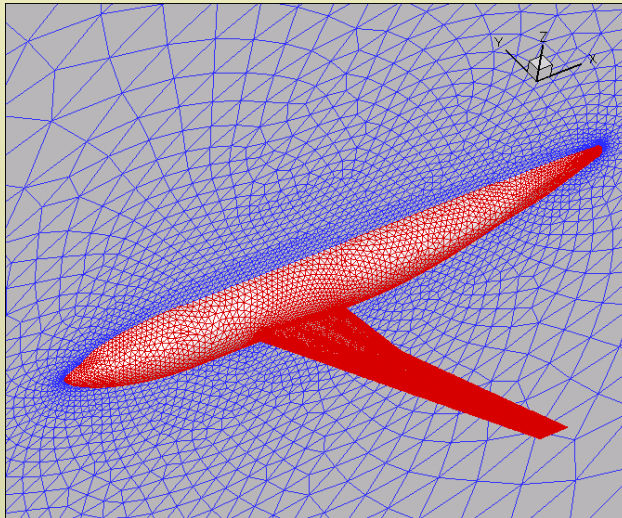
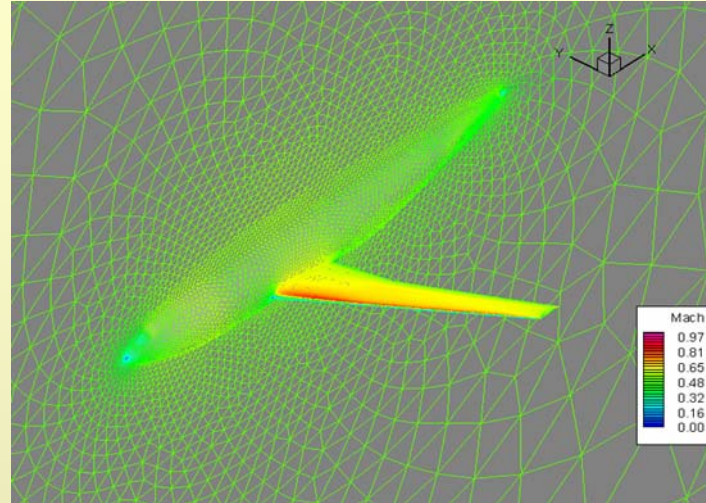
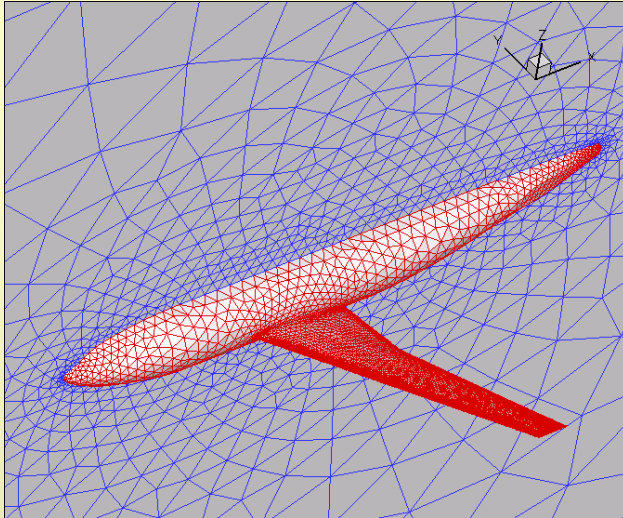
- [D] is Jacobian approximation
- Non-linear element-Jacobi (NEJ)

$$\boxed{\Delta \mathbf{U}_p^{n+1} = \left[\mathbf{D}_p^n \right]^{-1} \left(\mathbf{S}_p - \mathbf{R}_p(\mathbf{U}_p^n) \right)}$$



p	Size of [D]
1	5 x 5
2	20 x 20
3	50 x 50
4	100 x 100
5	280 x 280
6	420 x 420

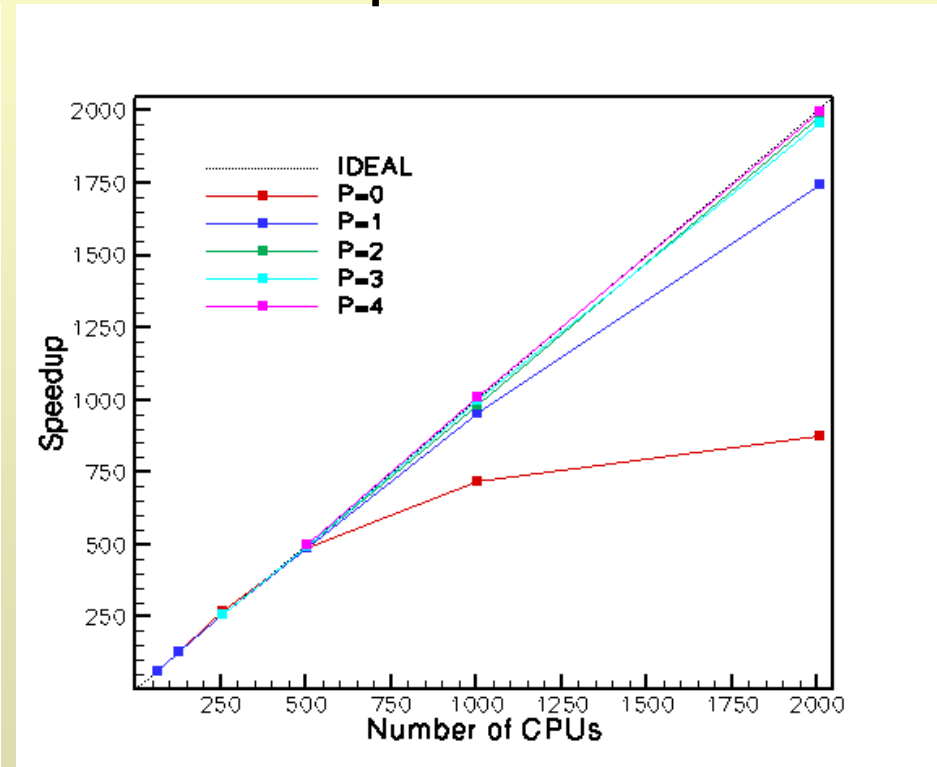
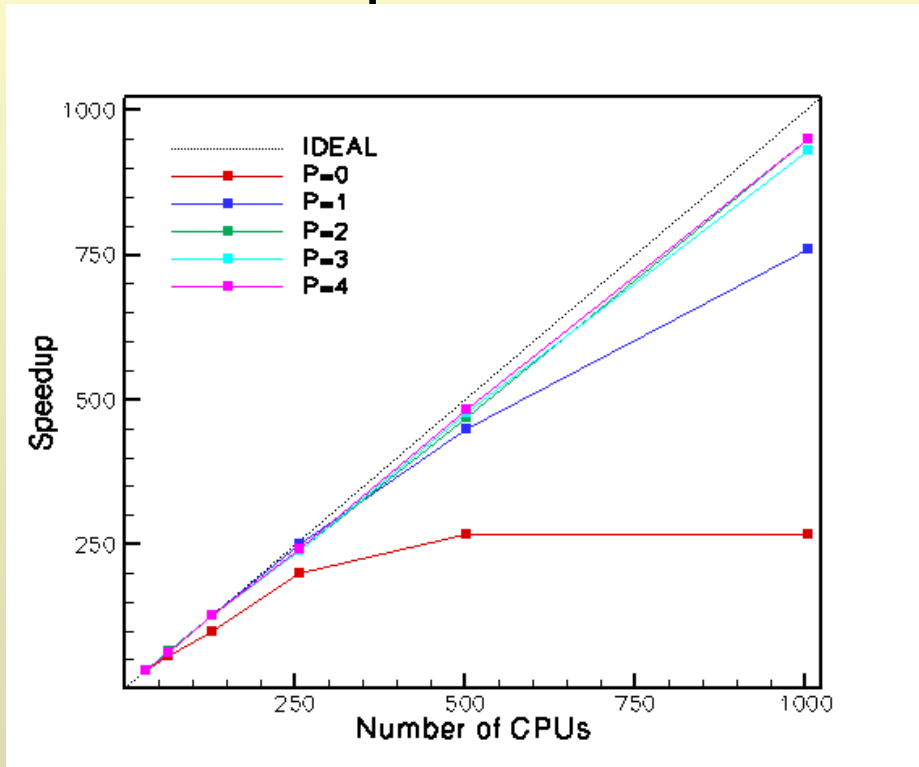
3D High-Order DG Results (Inviscid)



Parallel Performance: Speedup (1 MG-cycle)

185K pt mesh

2.5M pt mesh



- $p=0$ does not scale
- $p=1$ scales up to 500 proc.
- $p>1$ scales almost optimal

- $p=0$ does not scale
- $p=1$ scales up to 1000 proc.
- $p>1$ ideal scalability

Conclusions

- Current unstructured mesh solver technology is close to structured mesh solver capabilities
- All methods suffer from poorly quantified error sources
 - Particularly spatial discretization error
- Path forward will require new technology rooted in applied math
 - Sensitivity analysis techniques
 - Higher-order methods