

# Towards Automated Mesh Adaptation Using Simplex Cut Cells

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# Outline

- 1 Introduction
- 2 Discretization
- 3 Output-Based Adaptation
- 4 Cut Cells in Two Dimensions
- 5 Cut Cells in Three Dimensions
- 6 Research Directions
- 7 Conclusions



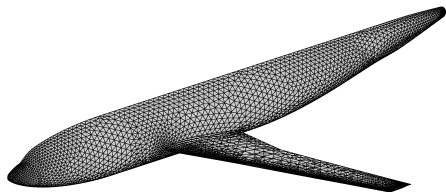
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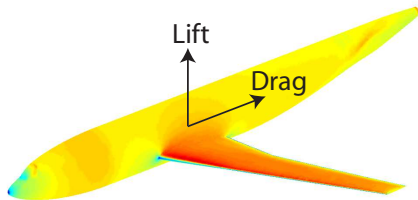


## CFD in aerospace engineering

- Actively used in design and analysis
- Supplements/replaces expensive wind-tunnel tests
- Reduces design cycle time and allows for innovative designs



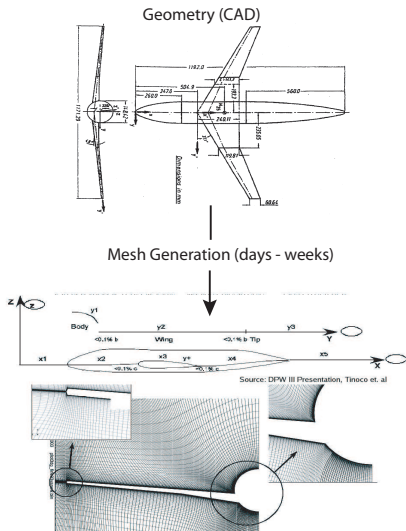
Geometry



Flow solution



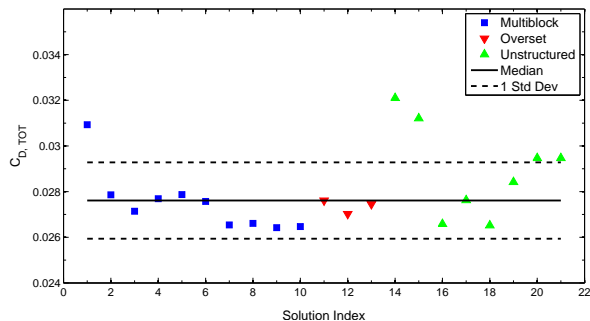
# Typical CFD Analysis



- Not representative of all CFD methods (e.g. Cartesian or boundary-potential methods)
- Typical use of finite volume in industry

## AAA Drag Prediction Workshop III – 2006

- Wing-body geometry,  $M = 0.75$ ,  $C_L = 0.5$ ,  $Re = 5 \times 10^6$
- Run on today's state-of-the-art CFD codes



### Differences in:

- Physical models
- Discretization
- **Mesh size distribution**

1 drag count ( $.0001 C_D$ )  $\approx$  4-8 passengers (Boeing 747-400)



# Identification of Problem

## Current CFD Practices:

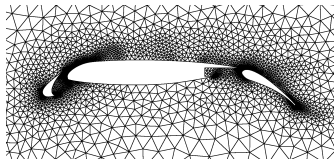
- Risk of unacceptably large errors is high
- Heavy “person-in-the-loop” involvement is required, especially during mesh generation
- Difficult to apply solution-based adaptation and optimization

## Key Problems:

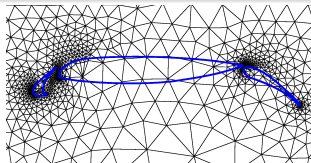
- Insufficient robustness
- Insufficient automation



## 1. Simplex cut-cell meshing for *high-order* solutions

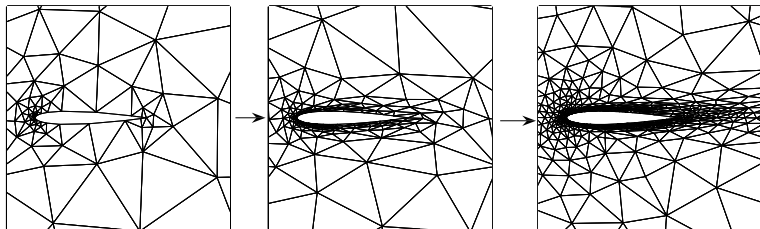


Boundary-conforming mesh



Simplex cut-cell mesh

## 2. Output-based *anisotropic* mesh adaptation for *high-order* solutions







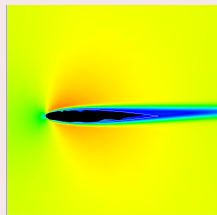
# Example: Adaptation + Cut Cells

User

- $M_\infty = 0.5$ ,  $Re = 5000$
- $C_D$  to within 1 count

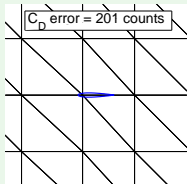


NACA 0012 geometry

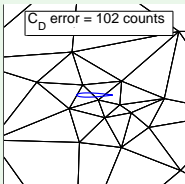


Automated

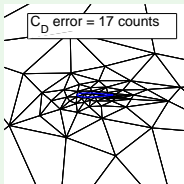
$C_D$  error = 201 counts



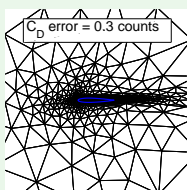
$C_D$  error = 102 counts



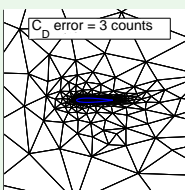
$C_D$  error = 17 counts



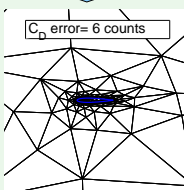
$C_D$  error = 0.3 counts



$C_D$  error = 3 counts



$C_D$  error = 6 counts



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# Compressible Navier-Stokes Equations

$$\frac{\partial \mathbf{u}}{\partial t} + \nabla \cdot \underbrace{\mathcal{F}_i(\mathbf{u})}_{\text{Inviscid Flux}} - \nabla \cdot \underbrace{\mathcal{F}_v(\mathbf{u}, \nabla \mathbf{u})}_{\text{Viscous Flux}} = 0$$

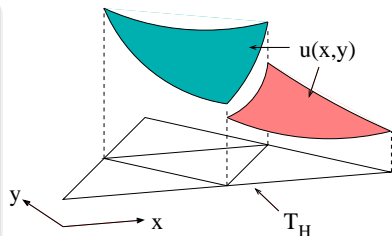
- 5 conservative variables:  $\mathbf{u} = [\rho, \rho u, \rho v, \rho w, \rho E]$
- 5 equations (conservation laws)
- Fluxes are nonlinear functions of  $\mathbf{u}$
- Interested primarily in steady-state ( $\partial \mathbf{u} / \partial t = 0$ )



# Discontinuous Galerkin Discretization

## High-order finite-element method:

- Solution/test space:  $\mathcal{V}_H = [V_H^p]^5$ ,  
 $V_H^p = \{v \in L^2(\Omega) : v|_\kappa \in P^p(\kappa) : \forall \kappa \in \mathcal{T}_H\}$
- Roe inviscid flux; 2<sup>nd</sup> form of Bassi and Rebay for elliptic term
- Discrete semi-linear form:  
 $\mathcal{R}_H(\mathbf{u}_H, \mathbf{v}_H) = 0, \quad \forall \mathbf{v}_H \in \mathcal{V}_H$



## Solution

- Newton GMRES
- Store full linearization
- Initial approximate time stepping

## Motivating features:

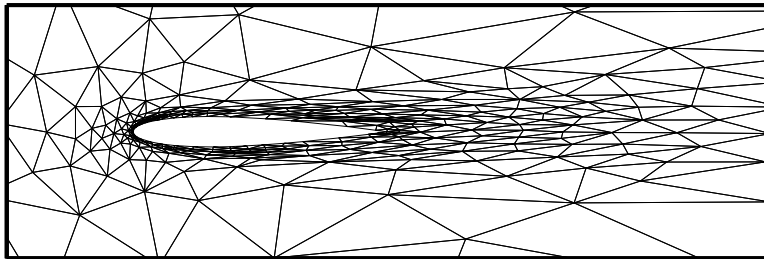
- High-order accuracy
- Element-wise compact stencil
- Ease of implementation

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# Output-Based Adaptation



$$C_D = 565.7 \text{ counts}$$

- How accurate is this value?
- Where is more resolution necessary to improve the accuracy?
- How should that resolution be added?

## Implementation

### 1. Output error estimation and localization

- $\mathcal{J}(\mathbf{u})$  = output of interest (lift, drag, etc.)
- $\mathbf{u}_H \in \mathcal{V}_H$  = approximate solution
- $\mathcal{J}(\mathbf{u}_H) - \mathcal{J}(\mathbf{u})$  = output error
- Solve for *adjoint*,  $\psi$  to estimate and localize the output error

### 2. Automated anisotropic $h$ -adaptation

- Anisotropy detection via extension of Hessian analysis to  $p > 1$
- Goal-oriented mesh optimization
- Re-meshing at every adaptation iteration





# Output Error Estimation: Local Error Indicator

## Extensive previous work:

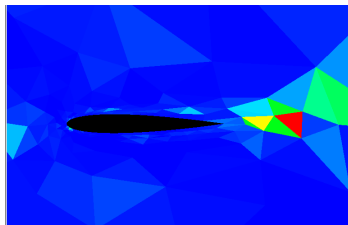
Pierce+Giles+Suli (2000),

Becker+Rannacher (2001),

Hartmann+Houston (2002),

Barth+Larson (2002)

Minor implementation differences



Error indicator for viscous case

$$\mathcal{J}(\mathbf{u}) - \mathcal{J}(\mathbf{u}_H) \approx \underbrace{\mathcal{R}_H(\mathbf{u}_H, \psi - \psi_H)}_{\text{Primal Residual}} \approx \underbrace{\mathcal{R}_H^\psi(\mathbf{u}_H; \mathbf{u} - \mathbf{u}_H, \psi_H)}_{\text{Adjoint Residual}}$$

$\mathbf{u} - \mathbf{u}_H$  and  $\psi - \psi_H$  estimated via reconstruction on enriched space.

## Elemental Error Indicator:

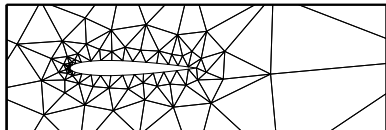
$$\epsilon_\kappa = \frac{1}{2} \left( \left| \mathcal{R}_h(\mathbf{u}_H, (\psi_h - \psi_H)|_\kappa) \right| + \left| \mathcal{R}_h^\psi(\mathbf{u}_H; (\mathbf{u}_h - \mathbf{u}_H)|_\kappa, \psi_H) \right| \right)$$



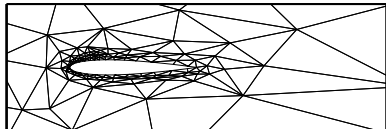
# Anisotropic Adaptation

**Idea:** refine elements with high error; coarsen elements with low error

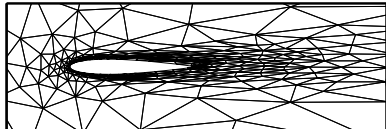
Iteration 0



Iteration 2



Iteration 4



- Use *a priori* output error estimate to relate element error to size request:

$$\epsilon_{\kappa} \sim h_{\kappa}^r$$

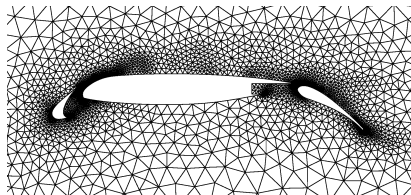
- Detect anisotropy by measuring  $p + 1$ st order derivatives of a scalar quantity (Mach number)
- Optimize mesh size to meet requested tolerance and to satisfy error equidistribution
- Meshing: BAMG in 2D, TetGen in 3D
- *Left:* NACA 0012,  $M = 0.5$ ,  $Re = 5000$ ,  $p = 2$  adapted on drag

# Outline

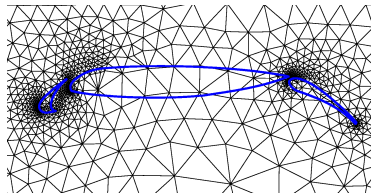
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# What Are Cut Cells?



Boundary-conforming mesh



Simplex cut-cell mesh

## Features

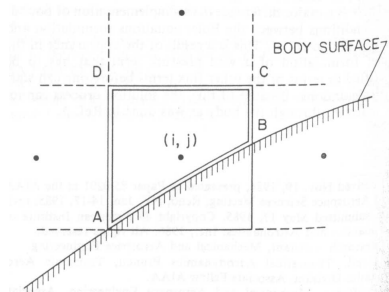
- Cut-cell meshes do not conform to geometry boundary
- Solution only exists inside the computational domain
- Premise: metric-driven meshing of a simple convex volume (e.g. box) is straightforward

# History: First Cut Cells

- 1979 – Purvis and Burkhalter: FV for 2D Full Potential Equations
- 1986 – Clarke, Salas, and Hassan: FV for 2D Euler
- 1987 – Gaffney, Salas, and Hassan: FV for 3D Euler

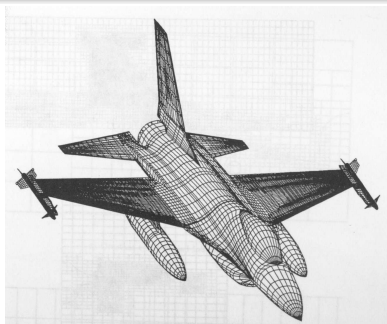
## Features

- Linear cut cells
- Agglomeration to remove small cells
- Uniform grids

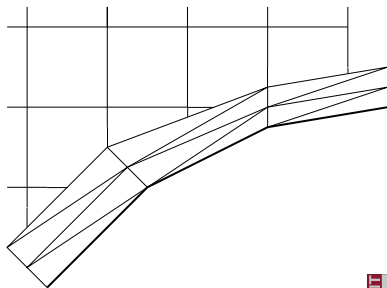


# History: Use in Industry Codes

- 1986 – Boeing’s TRANAIR: FEM for 3D Full Potential Equations; adaptation on geometry, user input, and solution; integration via Stoke’s theorem. Still in use today.
- 1995 – Karman’s SPLITFLOW (Lockheed): 3D RANS; required prismatic boundary layer mesh; outer flow via Cartesian cut cells.



TRANAIR

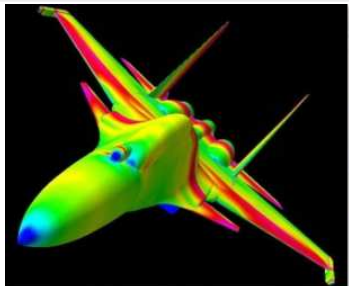


SPLITFLOW

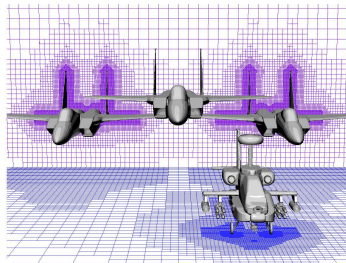


# History: Recent Work

- 1991 to present – MGAERO by Analytical Methods, Inc: finite difference for 3D Euler; multigrid, uniform grids.
- 1993+ – Application of adaptive refinement to Cartesian method for Euler; DeZeeuw, Powell, Coirier.
- 1999 to present – Cart3D: Mike Aftosmis *et al* , NASA; finite volume for 3D Euler; adaptively refined grids.



MGAERO



Cart3d



# Why Simplex Cut Cells?

**Objective:** A **robust, automated** mesher and **efficient** meshes

## Cartesian cut-cell method

- **Robust and automated grid generation**
- **Inability to adapt anisotropically**

## Simplex (triangles, tetrahedra) cut-cell method

- **Robust and automated grid generation**
- **Ability to adapt anisotropically in any direction**
- **Not as lean as Cartesian method**

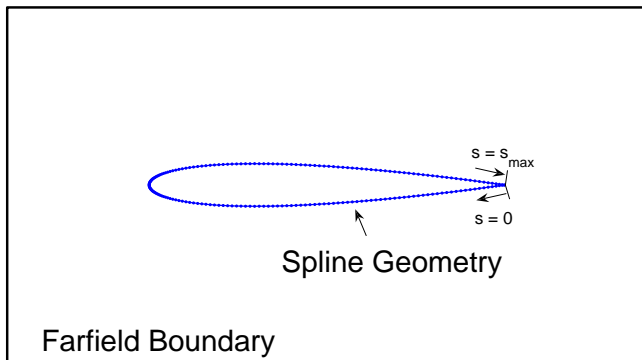




# 2D Geometry Definition

## Cubic splines

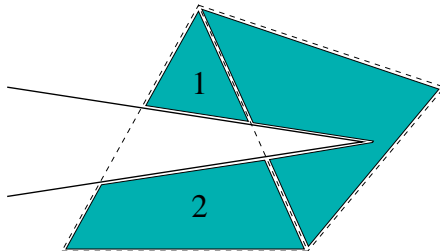
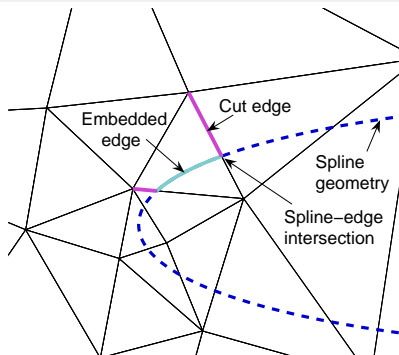
- Efficient treatment of curved boundaries
- Slope and curvature continuity at spline knots



# Intersection Problem

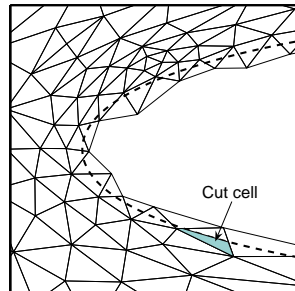
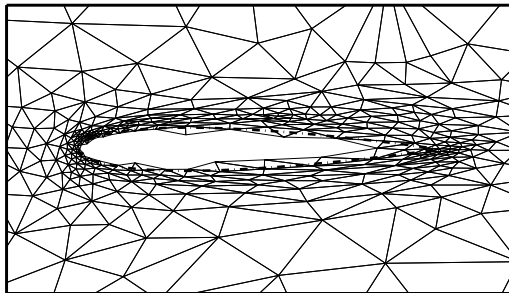
## Implementation

- Analytic intersections between splines and edges: cubic equation
- Multiply-cut triangles treated as a separate cut cells



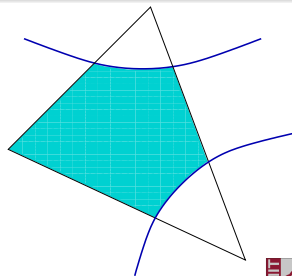
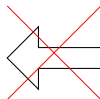
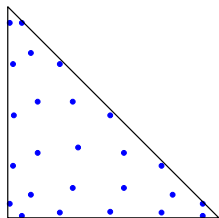
## Intersection Problem (ctd.)

- Triangles completely contained inside geometry removed from mesh structure
- Integration rules on cut cells/edges calculated in preprocessing



# Integration

- High-order finite element method requires integration over:
  - **Element boundaries** (edges in 2D, faces in 3D)
  - **Element interiors** (areas in 2D, volumes in 3D)
- Regular triangles and tetrahedra can be mapped to reference elements, where optimal integration rules exist
- These rules do not (in general) apply to cut cells, where areas and volumes are of irregular shape

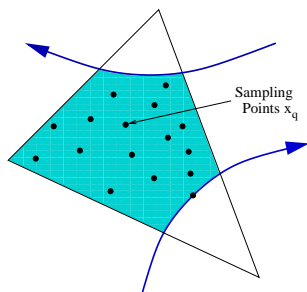


# Area Integration

## Goal

Sampling points,  $\mathbf{x}_q$ , and weights,  $w_q$  for integrating arbitrary  $f(\mathbf{x})$  to a desired order:

$$\int_{\kappa} f(\mathbf{x}) d\mathbf{x} \approx \sum_q w_q f(\mathbf{x}_q)$$



## Key Idea

Project  $f(\mathbf{x})$  onto space of high-order basis functions,  $\zeta_i(\mathbf{x})$ :

$$f(\mathbf{x}) \approx \sum_i F_i \zeta_i(\mathbf{x})$$

Choose  $\zeta_i(\mathbf{x})$  to allow for simple computation of  $\int_{\kappa} \zeta_i(\mathbf{x}) d\mathbf{x}$ .

## Area Integration (ctd.)

Set  $\zeta_i \equiv \nabla \cdot \mathbf{G}_i$  and use the divergence theorem:

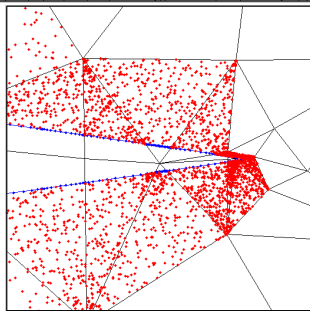
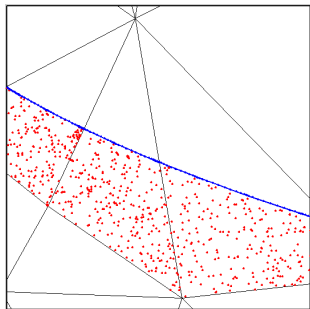
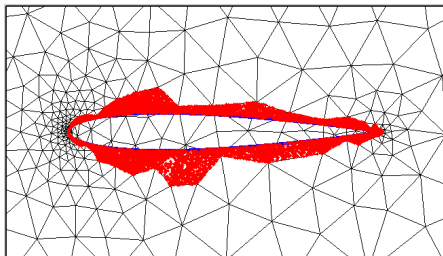
$$\int_{\kappa} \zeta_i d\mathbf{x} = \int_{\kappa} \nabla \cdot \mathbf{G}_i d\mathbf{x} = \int_{\partial\kappa} \mathbf{G}_i \cdot \mathbf{n} ds$$

- $\mathbf{G}_i$  = a standard high-order basis (e.g. tensor product)
  - Line integrals over  $\partial\kappa$  using 1D edge formulas
- 
- Projection  $f(\mathbf{x}) \approx \sum_i F_i \zeta_i(\mathbf{x})$  minimizes the least-squares error at randomly-chosen sampling points,  $\mathbf{x}_q$ , inside the cut cell
  - QR factorization and integration over  $\kappa$  leads to an expression for the quadrature weights:

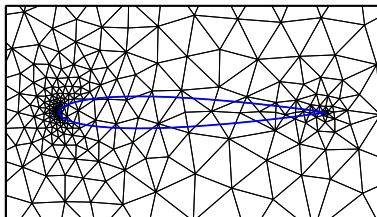
$$\int_{\kappa} f(\mathbf{x}) d\mathbf{x} \approx \sum_i F_i \int_{\kappa} \zeta_i(\mathbf{x}) d\mathbf{x} = \sum_q f(\mathbf{x}_q) \underbrace{Q_{qj} (R^{-T})_{ji}}_{W_q} \int_{\kappa} \zeta_i(\mathbf{x}) d\mathbf{x}$$

# Example: Quadrature Points

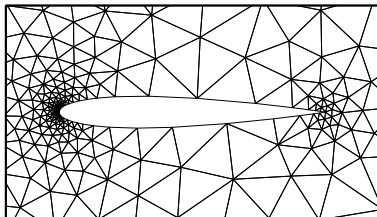
- NACA 0012
- 12 Gauss points per cut edge and spline segment
- Over 200 interior sampling points per element



# Example: Flow Solution



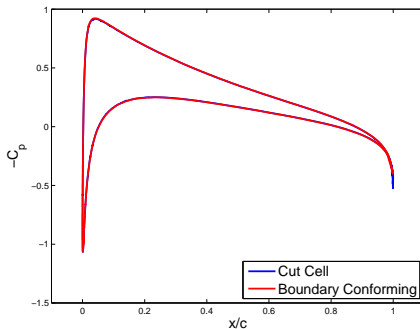
Cut-cell mesh



Boundary-conforming mesh

- $M = 0.5$ ,  $\alpha = 3^\circ$
- $p = 2$  interpolation

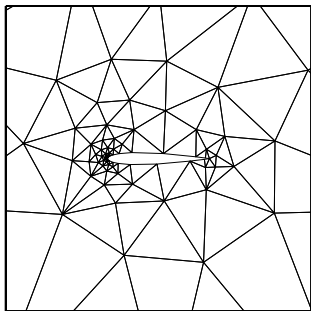
$C_p$  comparison



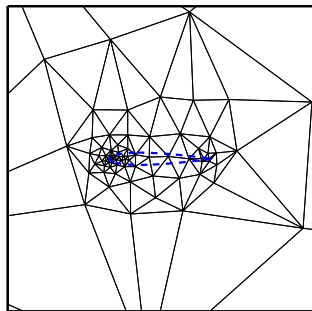


# Drag Adaptation in a Viscous Case

NACA 0012,  $M = 0.5$ ,  $Re = 5000$ ,  $\alpha = 2^\circ$



Initial boundary-conforming  
mesh

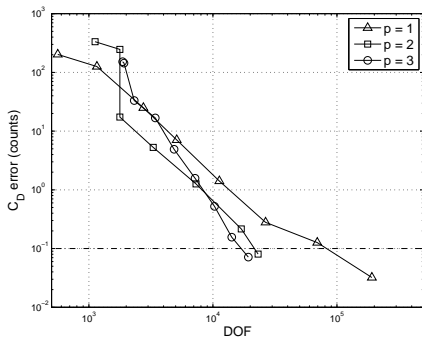


Initial cut-cell mesh

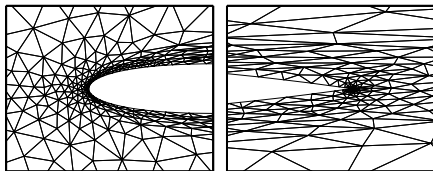
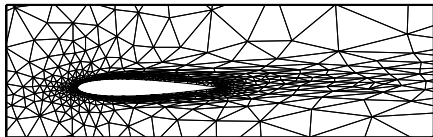


# Viscous Case: Error Convergence

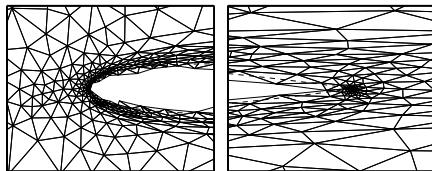
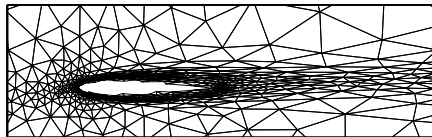
- Degree of freedom (DOF) vs. drag output error for  $p = 1, 2, 3$ .
- Requested tolerance is 0.1 drag counts (horizontal line).
- Cut-cell and boundary-conforming results are similar.



# Viscous Case: Final Meshes



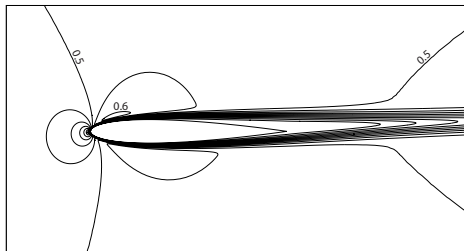
$p = 3$  adapted boundary-conforming mesh



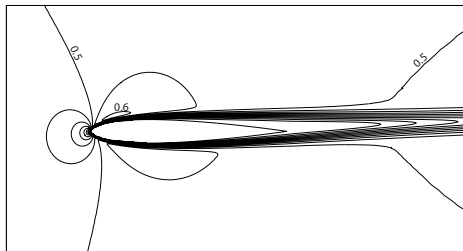
$p = 3$  adapted cut-cell mesh

$p = 1$  meshes have approximately 50 times more elements

# Viscous Case: Mach Number Contours



Boundary-conforming,  $p = 3$



Cut-cell,  $p = 3$

$p = 1$  and  $p = 2$  contour lines are very similar



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# Extension to Three Dimensions

- Cut-cell mesh generation becomes more difficult:
  - Geometry representation is not as straightforward as in 2D
  - Harder intersection problem: volume-surface instead of area-line
  - Integration rules needed on geometry surface, cut faces, and cut elements
- However, generating 3D boundary-conforming meshes is much more difficult compared to 2D:
  - Meshing around intricate 3D geometries is not trivial
  - No robust automated technique for curved geometries



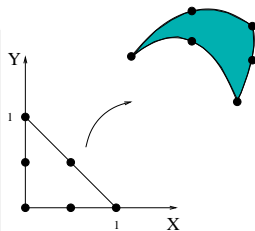
# Geometry Definition

- Quadratic patches, 6 nodes per patch
- Patch surface ( $\mathbf{x}$ ) is given analytically:

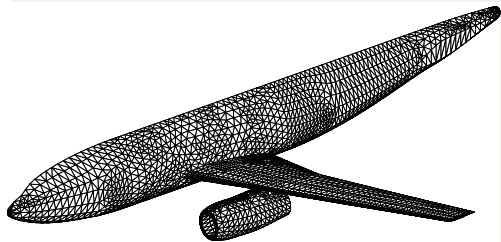
$$\mathbf{x} = \sum_j \phi(\mathbf{X})_j \mathbf{x}_j,$$

$\mathbf{X} = [X, Y]$ : patch ref coords

- Water-tight representation (no holes)

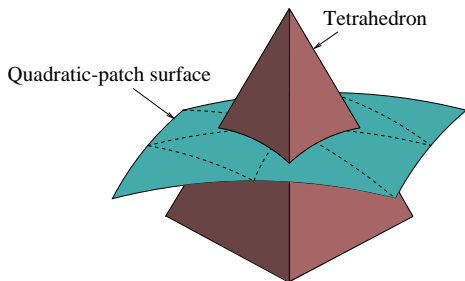


Patch reference space

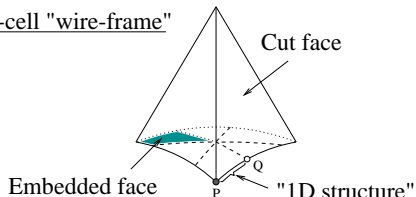


- Not exact; intermediate surface representation
- Surface tessellation and geometry interrogation from CAD via CAPRI
- Efficient resolution of curved surfaces

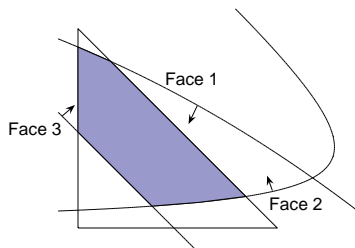
# Intersection with Background Mesh



## Cut-cell "wire-frame"



- Analytical intersection possible
- **Enabling feature:** intersection between a plane and a quadratic patch is a conic section (ellipse, hyperbola, etc.) in  $(X, Y)$
- Robustness of cutting algorithm relies on robustness of conic-conic intersections



Conics in patch reference space



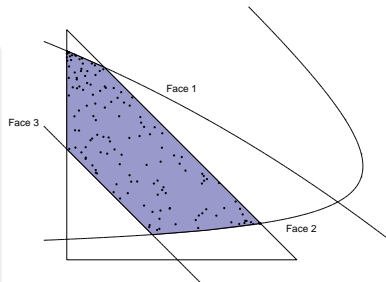


## Requirements

- 2D integration on embedded boundary faces (on patches)
- 2D integration on cut faces (from background tetrahedra)
- 3D integration on cut-cell interiors

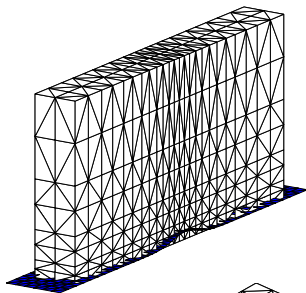
## Methodology

- Gauss points on 1D edges of 2D embedded and cut faces
- Sampling point speckling for face integration (as in 2D)
- 3D extension of point speckling for cut elements

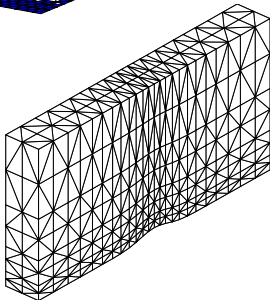


# Example: Flow Solution

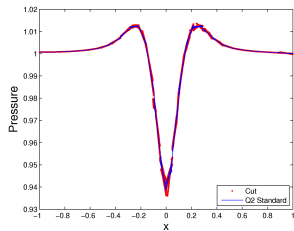
Cut-cell mesh:



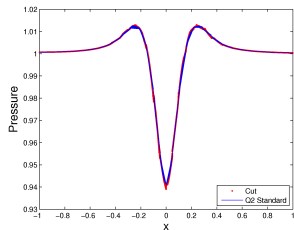
Boundary-conforming mesh:



$p = 1$

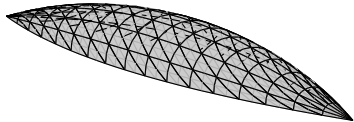


$p = 2$

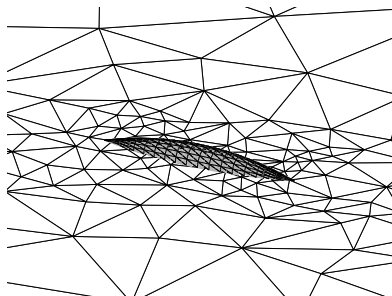


# Flow Around a Football

- Inviscid,  $M_\infty = 0.3$  flow around a body of revolution
- Model half the geometry



Surface representation: 256 quadratic patches

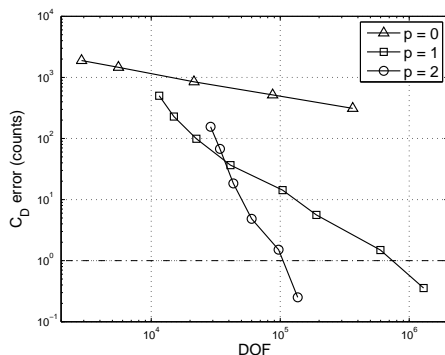


Initial background mesh: 2883 elements



# Football: Error Convergence

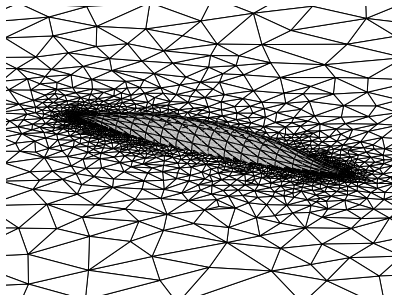
- Adapted on drag, with error tolerance of 1 drag count
- $C_D$  measured using frontal cross-sectional area



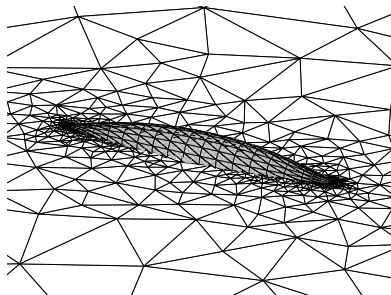
- $p=0$  is not practical for accurate computation
- $p=2$  converges much faster than  $p=1$



# Football: Adapted Meshes

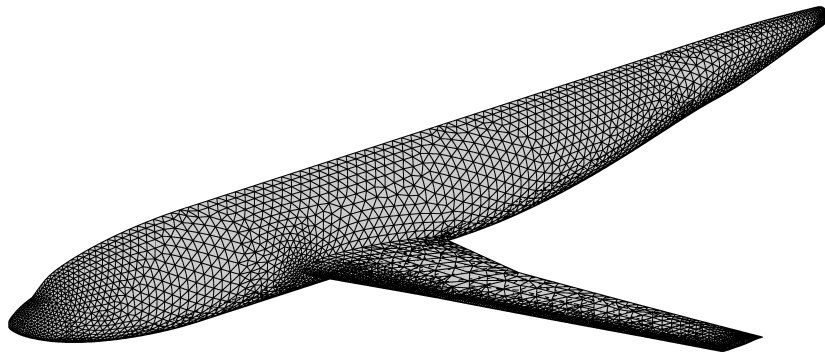


$p = 1$ : 66304 elements: error = 4.0 drag counts



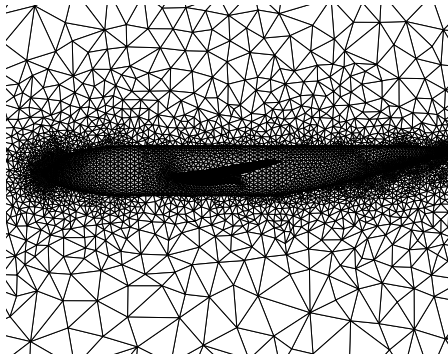
$p = 2$ : 11354 elements: error = 0.6 drag counts

# Wing-Body: Geometry

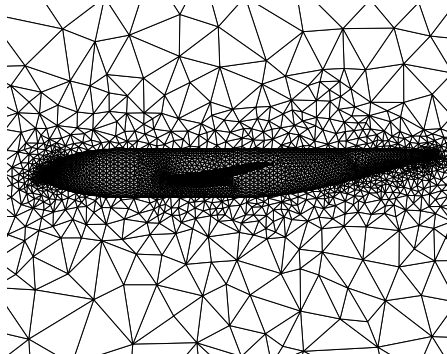


- Geometry from Drag Prediction Workshop
- 10,000 quadratic surface patches

# Wing-Body: Adapted Meshes

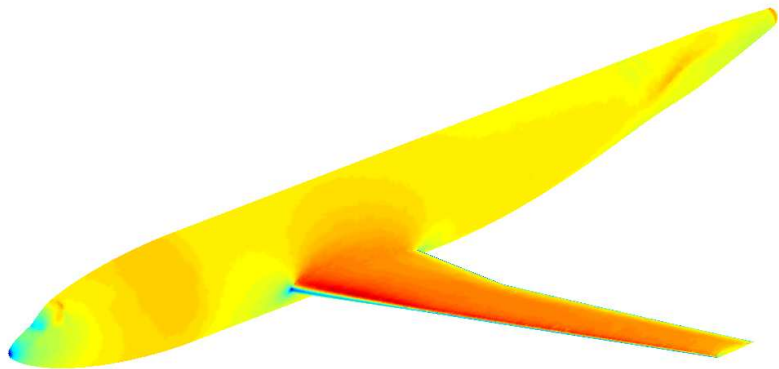


$p = 1$ : 300,000 elements



$p = 2$ : 85,000 elements

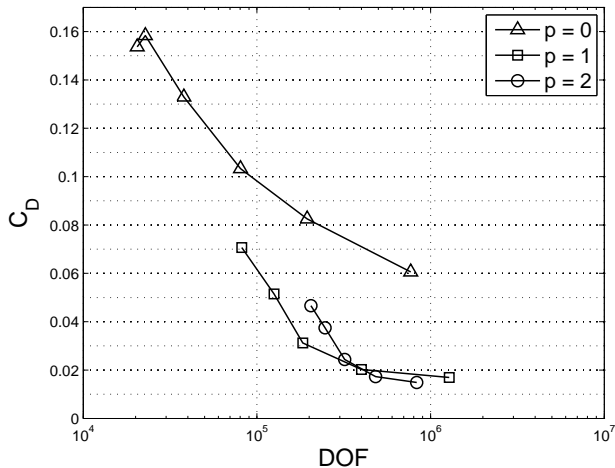
# Wing-Body: Solution



- Inviscid  $M_\infty = 0.1$  flow
- Mach number contours shown for a  $p = 2$  solution



# Wing-Body Drag Comparison



# Outline

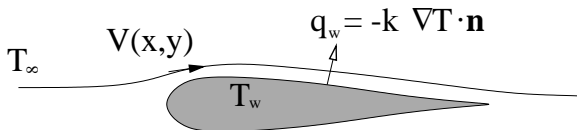
- 1 Introduction
- 2 Discretization
- 3 Output-Based Adaptation
- 4 Cut Cells in Two Dimensions
- 5 Cut Cells in Three Dimensions
- 6 Research Directions**
- 7 Conclusions



# High Peclet Number Convection-Diffusion

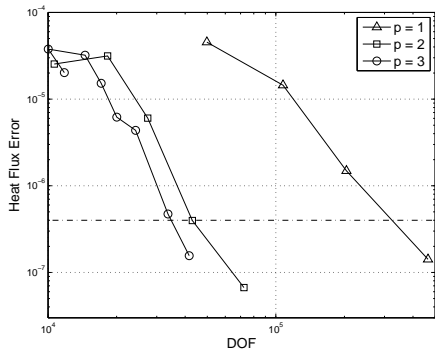
- Test effectiveness of cut-cells + adaptation for highly-anisotropic boundary layer meshes
- Thickness of boundary layer governed by Peclet number,  $Pe$

$$\nabla \cdot (\mathbf{V}T) - \nabla \cdot (k\nabla T) = 0, \quad Pe = \frac{V_\infty L}{k}$$



# $Pe = 4 \times 10^8$ : Error Convergence

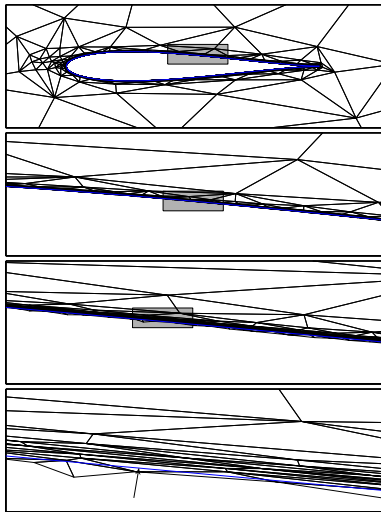
- $Pe = 4 \times 10^8$  simulates turbulent inner layer at  $Re \sim 10^6$
- Heat flux output:  $\mathcal{J} = \int_{\text{airfoil}} q_w ds$  + dual consistent terms
- Error tolerance is 1% of true heat flux



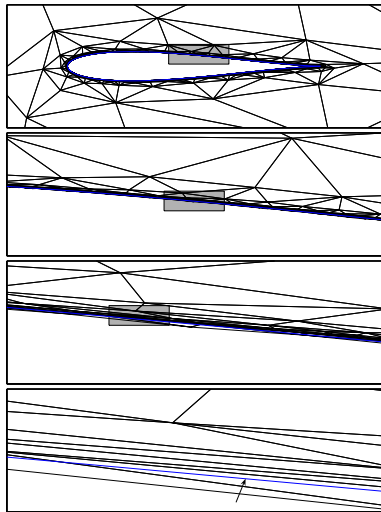
- $p = 1$  requires a factor of 10 more degrees of freedom than  $p = 3$
- $p = 2$  performance is similar to  $p = 3$



# $Pe = 4 \times 10^8$ : Adapted Meshes



$p = 2$

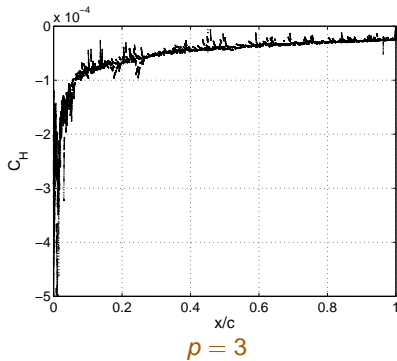
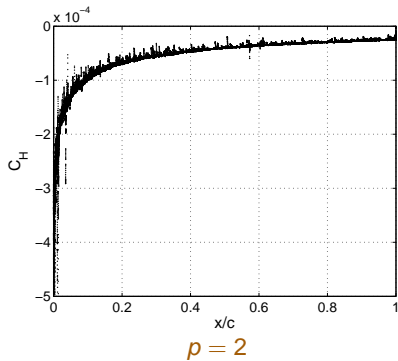


$p = 3$

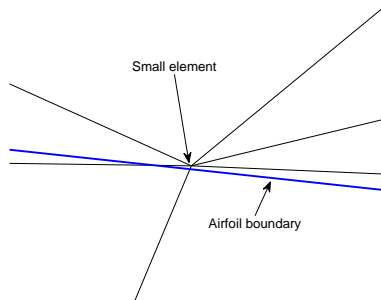
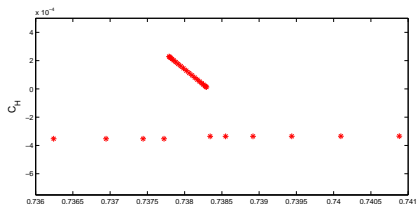


# $Pe = 4 \times 10^8$ : Heat Transfer Coefficient

$C_H = q_w / (V_\infty \Delta T)$  along airfoil surface



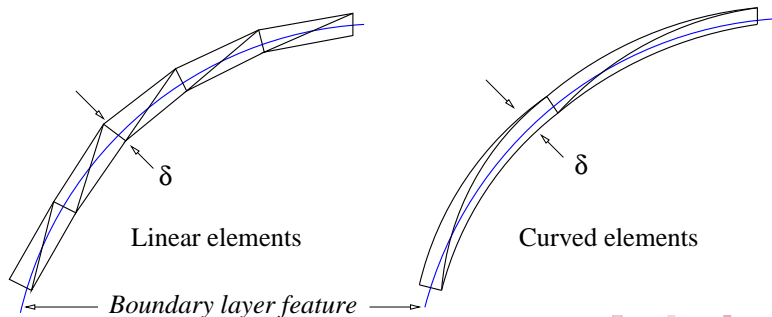
# Diffusion Discretization on Small Elements



- Noise in derivative quantities observed on small elements adjacent to large elements
- Not specific to cut-cells, but cut-cell meshes are likely to contain small elements
- Problem due to viscous discretization + under-resolution
- Possible solutions:
  - Merge very small elements with neighbors
  - Seek a more robust discretization

# Resolution of Curved Features

- Anisotropic features are efficiently resolved with anisotropic elements
- Feature curvature limits maximum element anisotropy when linear elements are used
- To take advantage of curved elements, solution representation must be in mapped (curved) space



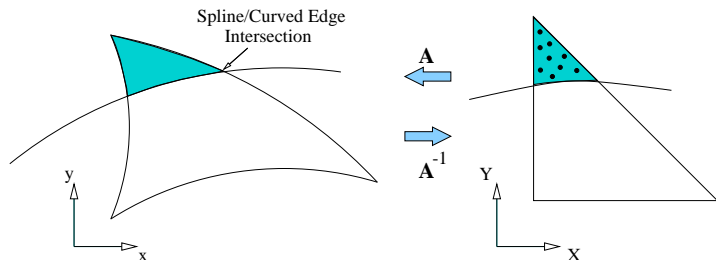


# Curved Elements

- DG boundary-conforming meshes employ curved elements to adequately represent curved boundaries
- Curved features may exist away from boundaries:
  - Unsteady shear layers
  - Curved shocks
- Ideally, elements should be curved based on the solution, not necessarily/just on the geometry
- Globally curving mesh elements while respecting geometry boundaries is a challenging task
- Cut cells with curved background meshes offer an alternative approach, more suitable for automation



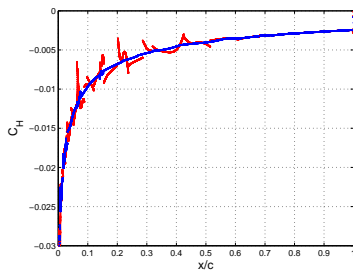
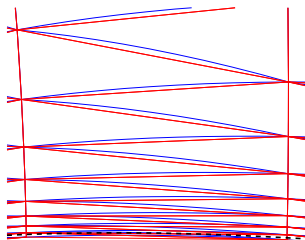
## 2D Cut Cells with Curved Elements



- Spline/curved-edge intersections obtained by applying Newton-Raphson method to the system of nonlinear equations
- Area integration rule derived in the element ref. space  $(X, Y)$ 
  - Solution is polynomial in  $(X, Y)$
  - Inverse of nonlinear mapping ( $A^{-1}$ ) is required to transform spline quadrature/intersection points into  $(X, Y)$  space
- Aside from cutting/integration, no fundamental code changes are required to incorporate curved cut cells

## 2D Cut Cells with Curved Elements (ctd.)

$Q = 1$  versus  $Q = 2$  boundary-layer cut-cell meshes



$p = 2$  solutions

- Convection diffusion for a Joukowski airfoil: output from cut  $Q = 2$  mesh shows marked improvement over cut  $Q = 1$  mesh
- Meshes were created manually - automated generation and adaptation of curved-element background meshes and extension to 3D is an ongoing research topic

# Outline

- 1 Introduction
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# Conclusions

- Cut-cells offer an automated alternative to boundary-conforming mesh generation, which is very often the bottleneck in CFD analysis and design
- Simplex cut cells allow for anisotropic meshes, which are necessary for practical viscous computations
- Adaptation with an output error estimator removes user guesswork from geometry-to-solution analysis process
- Curved background elements are more efficient at resolving curved anisotropic features; a robust adaptive scheme needs to be developed to take advantage of this efficiency
- Further work is required in making the viscous discretization more well-behaved on small elements



# Acknowledgements

- Ph.D. Thesis advisor: David Darmofal
- Project-X team
- Department of Energy Computational Science Graduate Fellowship (DOE CSGF)



Questions?

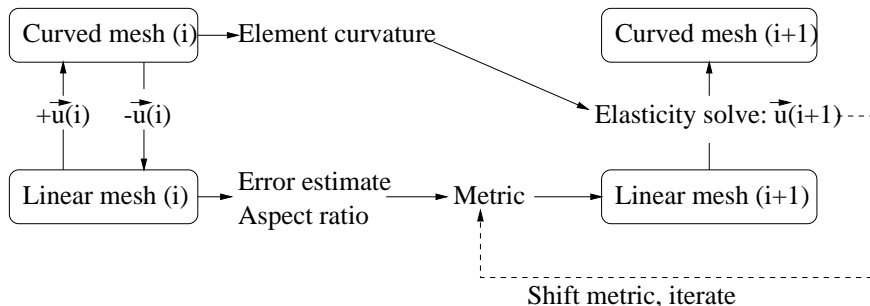


- **Decreasing computational cost of cut cells**
  - Reduce number of sampling points by more sophisticated (not random) selection; e.g. electric charge analogy
  - Allow for geometry approximation when background mesh is coarse
- **Improving conditioning of 3D volume integration rules at high orders**
  - Seek better support for integrand basis functions
  - Other polyhedra as alternatives to fitted bounding-box
- **Smoother geometry representation in 3D**
  - Investigate feasibility of intersection with more continuous geometry representations



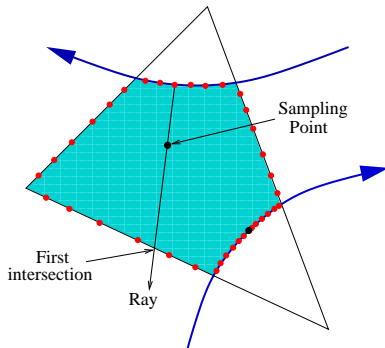


# Curved Mesh Adaptation



# Sampling Point Selection

- $\mathbf{x}_q$  must lie inside the cut cell to keep the integrand evaluations physical for non-linear problems
- Currently choosing  $\mathbf{x}_q$  randomly via ray-casting:



- Clusters of sampling points are undesirable in terms of QR conditioning  $\Rightarrow$  use *oversampling*

## Hessian Matrix

- Based on measuring degree and direction of quadratic variation.
- Standard practice in finite volume and linear FEM.
- Not reliably applicable to high-order solutions.

Example:  $u = 1.0 + (x^2 + 16y^2) + \epsilon(64x^3 + y^3)$ ,  $\epsilon \ll 1$

The Hessian matrix is

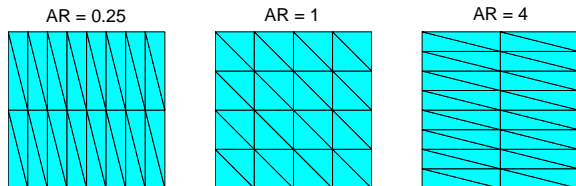
$$H = \begin{bmatrix} u_{xx} & u_{xy} \\ u_{yx} & u_{yy} \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 0 & 32 \end{bmatrix} + O(\epsilon)$$

Ignoring  $O(\epsilon)$  terms, Hessian analysis predicts

$$AR \equiv \frac{\Delta x}{\Delta y} = \sqrt{\frac{32}{2}} = 4.0.$$



# Role of Anisotropy (continued)



## $L_2$ Interpolation Error

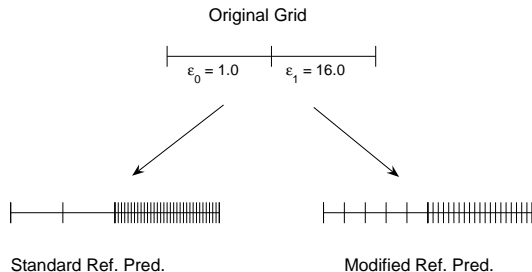
AR	$p = 1$	$p = 2$
0.25	2.31E-1	2.19E-7
1.0	5.91E-2	1.42E-6
4.0	2.37E-2	1.14E-5

## For high-order ( $p > 1$ ) anisotropy measures

- Use direction and magnitudes of  $(p + 1)^{\text{st}}$  derivatives.
- Directions of min/max. H.O. derivatives no longer guaranteed to be orthogonal.
- Currently employ a brute-force search for max H.O. derivative.



# Mesh Optimization Algorithm: Example



Assumed  $\epsilon_{\kappa} \sim h_{\kappa}^1$

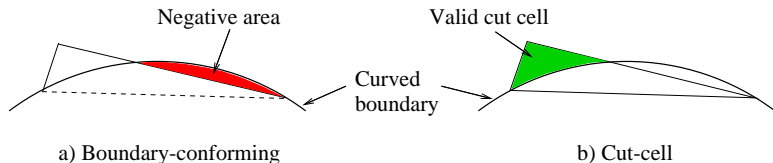
Standard Refinement Prediction : 34 elements  
Modified Refinement Prediction : 25 elements



# The Cut-Cell Advantage

## Boundary-conforming mesh generation

- Common bottleneck in geometry-to-solution process
- Difficult (not robust) for complex 3D geometries
- Prone to failure on curved boundaries



## Cut-cells

- Naturally handle curved boundaries and complex geometries
- Burden of robustness transferred to computational geometry
- Fully-automated mesh generation is possible