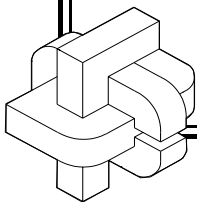


# Grids and Solutions from Residual Minimization

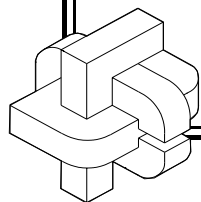
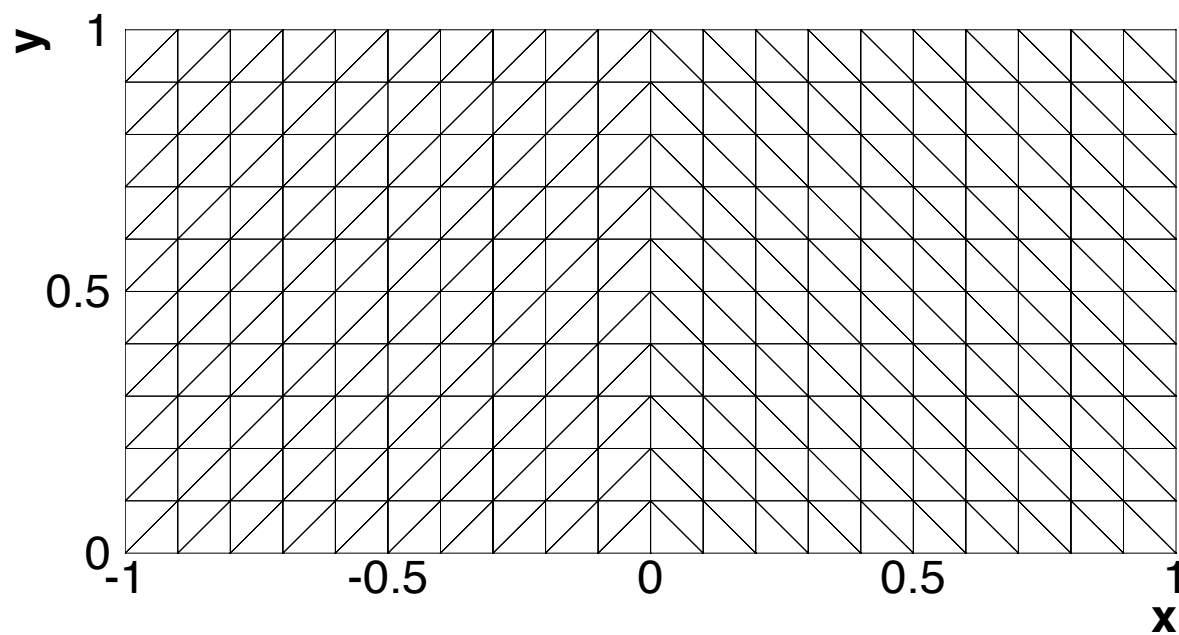
HIROAKI NISHIKAWA

*W.M.Keck Foundation Laboratory for Computational Fluid Dynamics  
Department of Aerospace Engineering, University of Michigan,  
Ann Arbor, Michigan 48109*



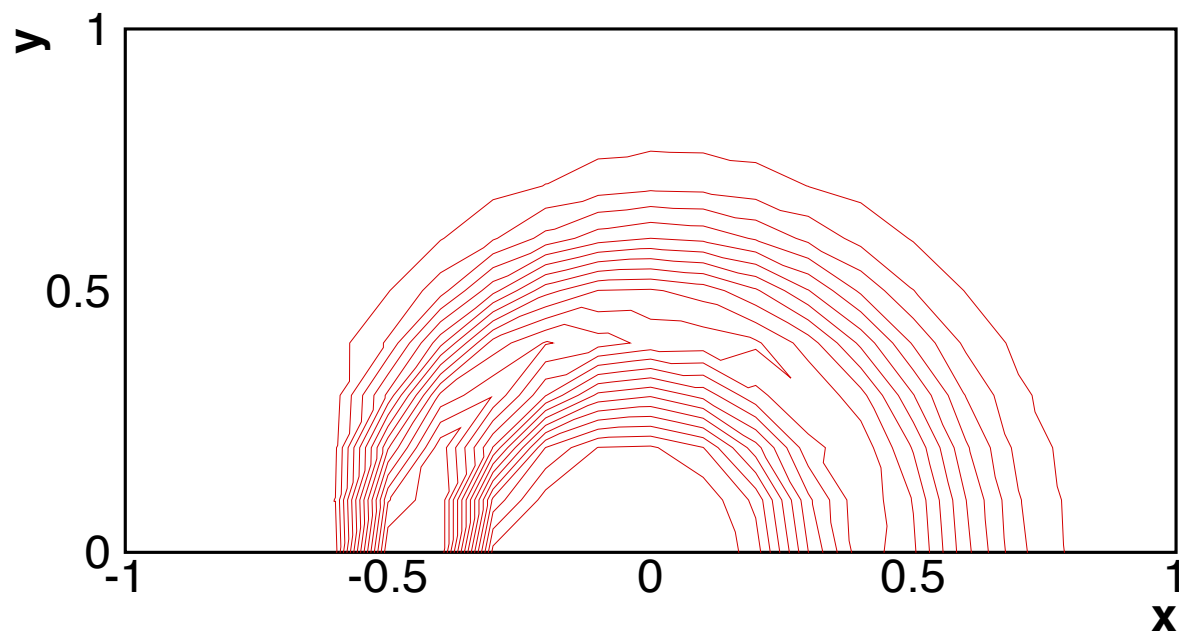
# Circular Advection

$$y \partial_x u - x \partial_y u = 0$$

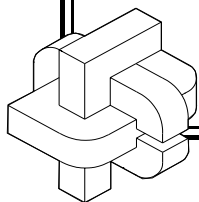


# Circular Advection

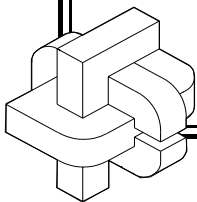
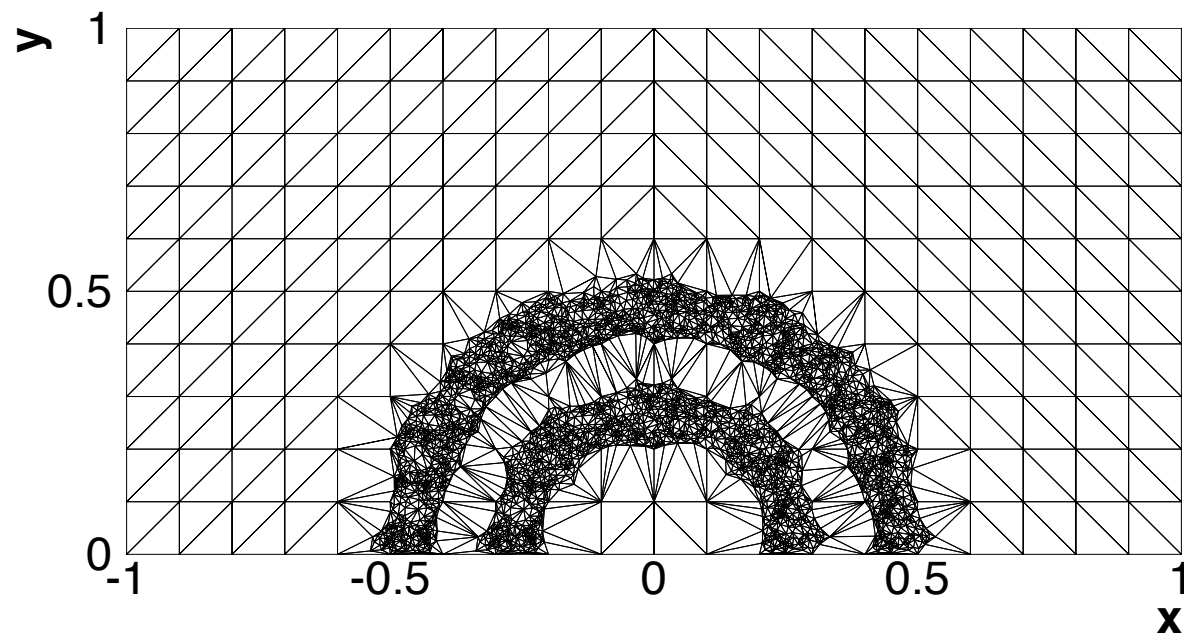
$$y \partial_x u - x \partial_y u = 0$$



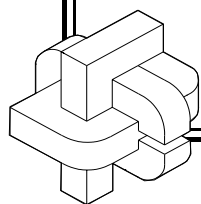
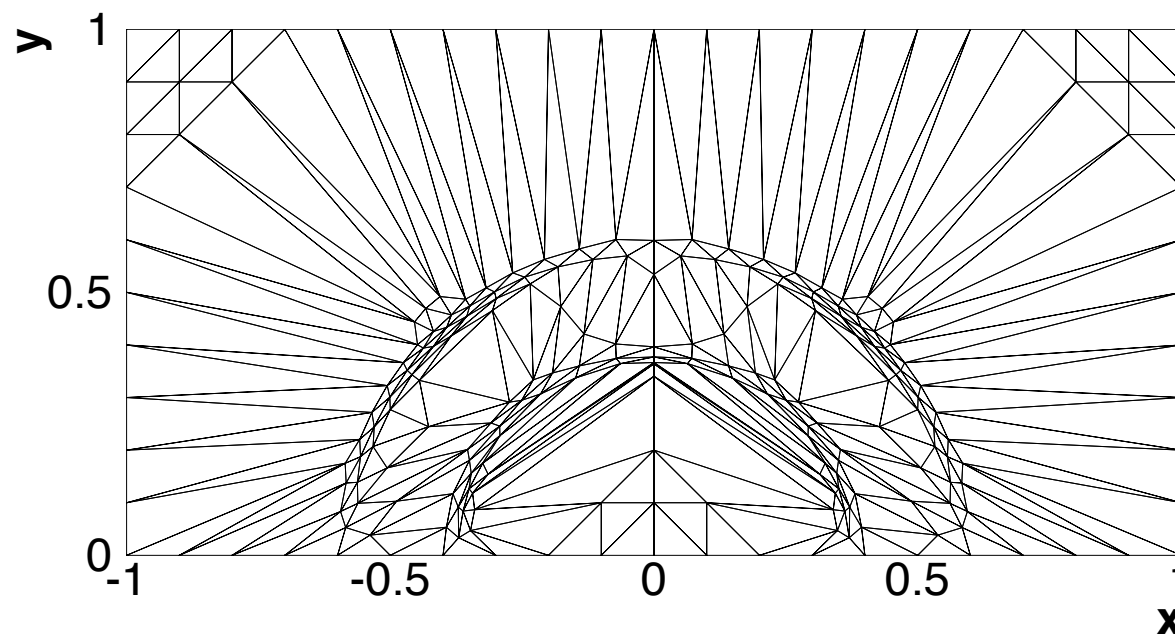
*A typical solution on the regular grid.*



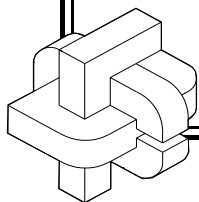
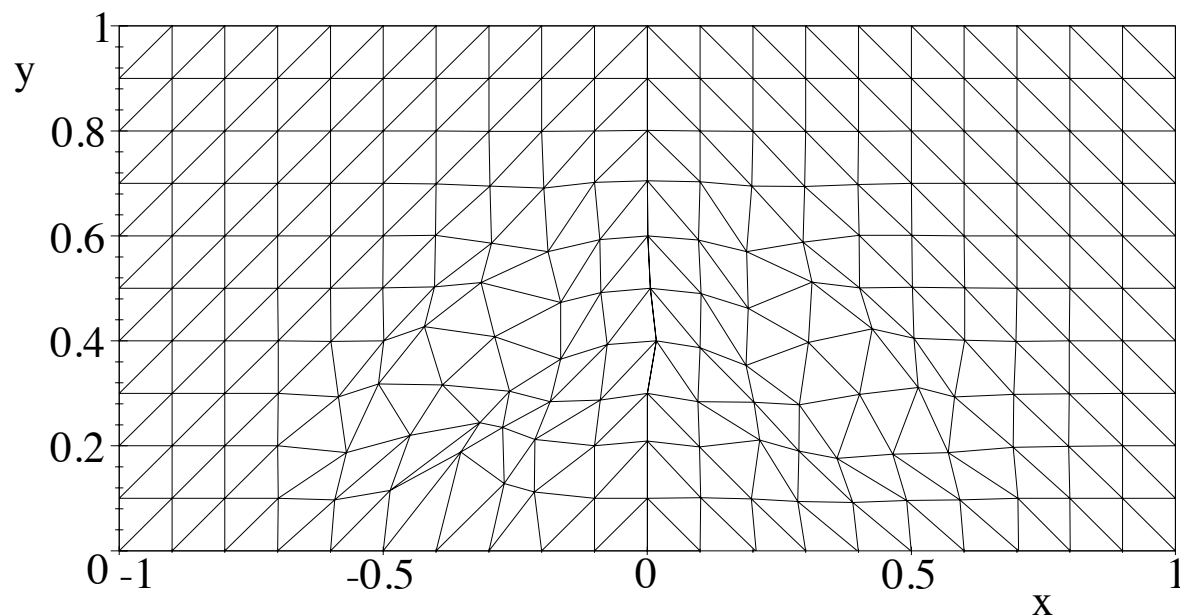
# Mesh Refinement



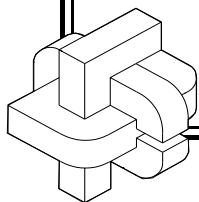
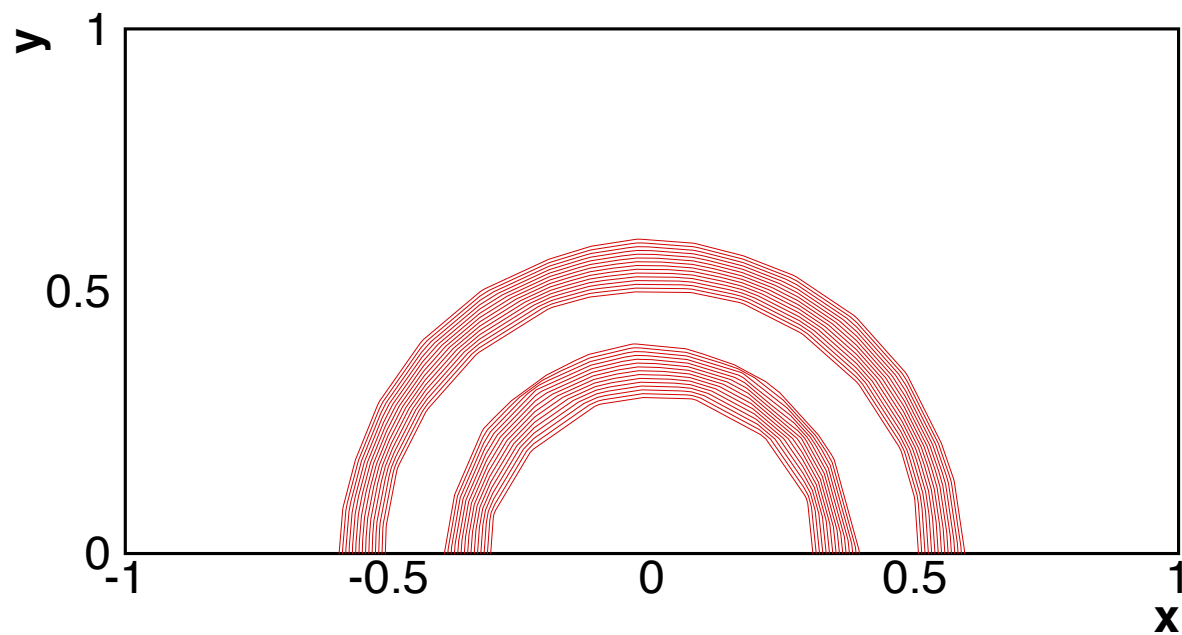
# Mesh Movement



# An Adaptive Grid



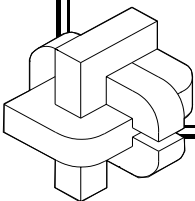
# An Adaptive Grid



# Why Triangular Grids?

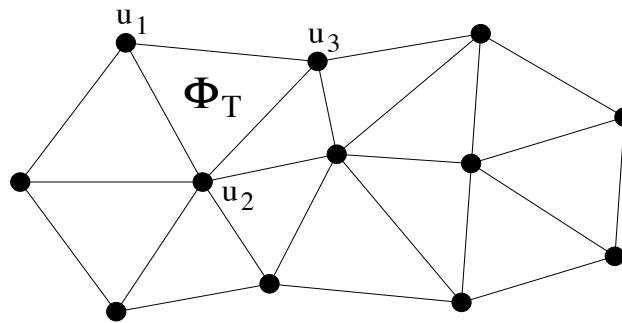
- Easy to Create for Complex Domains
- Easy to Insert/Delete Nodes
- Easy to Change the Connectivity (Edge Swapping)
- *Additional Degrees of Freedom*

The Number of Elements  $\approx 2 \times$  The Number of Nodes





# Least-Squares Residual Minimization



FLUCTUATION:

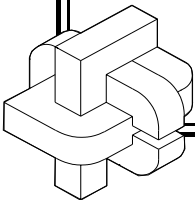
$$\Phi_T = \int_T [\partial_x \mathbf{F} + \partial_y \mathbf{G}] \, dx \, dy$$

NORM:

$$\mathcal{F} = \frac{1}{2} \sum \Phi_T^t Q_T \Phi_T$$

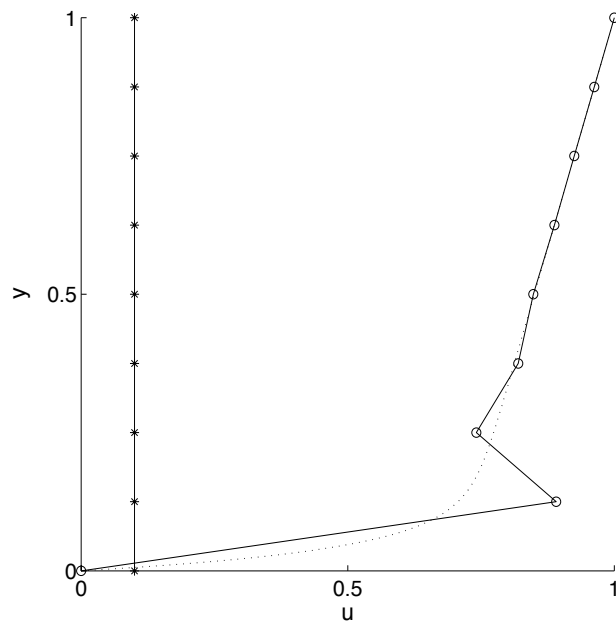
UPDATES:

$$\delta \mathbf{u}_j = -\omega_u \frac{\partial \mathcal{F}}{\partial \mathbf{u}_j}, \quad \delta \mathbf{x}_j = -\omega_x \frac{\partial \mathcal{F}}{\partial \mathbf{x}_j}$$

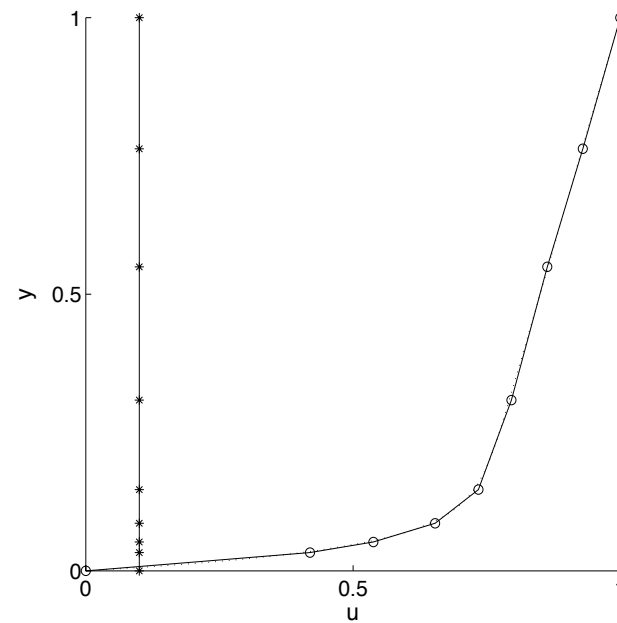


# Example 1: One-Dimensional Problem

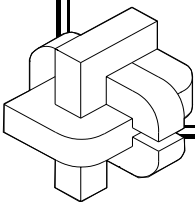
$$\epsilon \frac{d^2 u}{dy^2} + \frac{du}{dy} - a = 0$$



A Fixed Uniform Grid

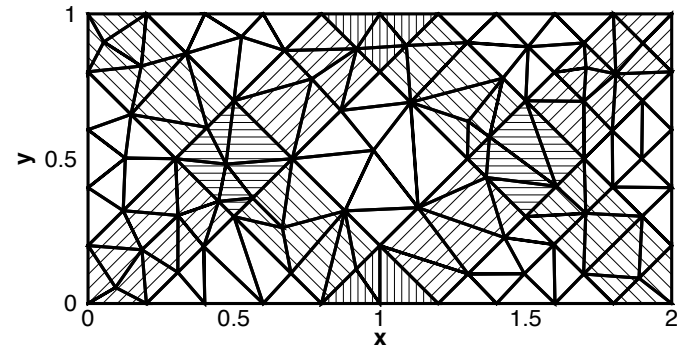
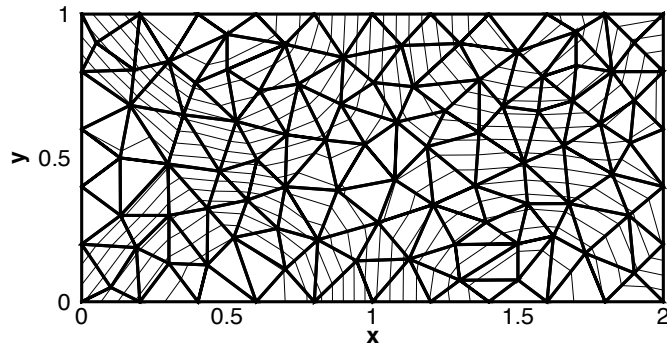


An Adaptive Grid

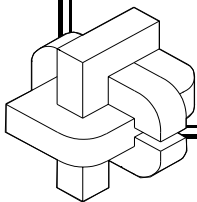


## Example 2: Hyperbolic Problem

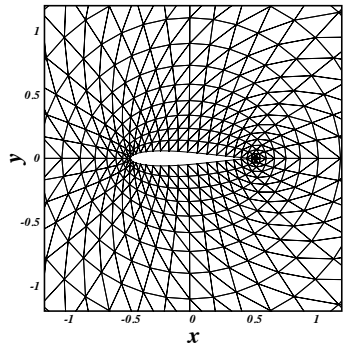
$$(1 - M_\infty^2)\partial_x u + \partial_y v = 0, \quad \partial_x v - \partial_y u = 0$$



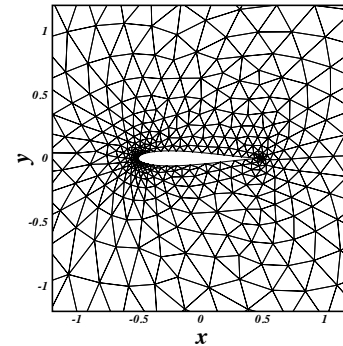
Supersonic flow through a duct,  $M_\infty = \sqrt{2}$  with a small incidence.



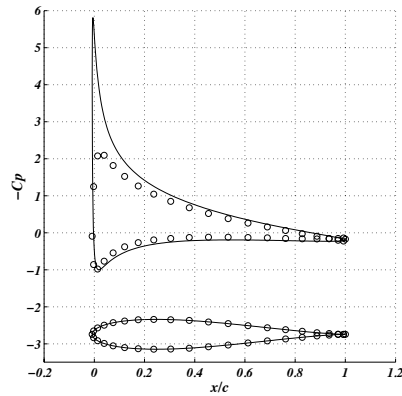
# Example 3: Elliptic Problem



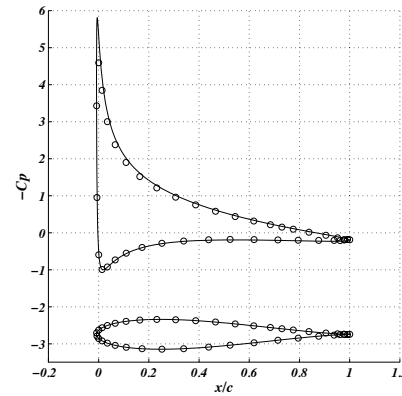
Regular 40x20 Grid



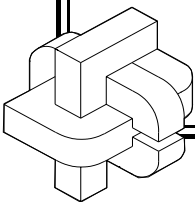
An Adaptive Grid



$C_p$  Distribution



$C_p$  Distribution



# Linear Advection

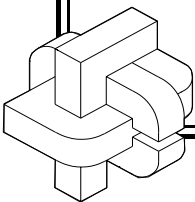
LINEAR ADVECTION EQUATION:

$$a \partial_x u + b \partial_y u = 0$$

*The solution  $u$  is convected in the direction  $(a, b)$ .*

CHARACTERISTIC EQUATION:

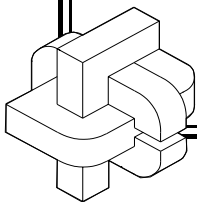
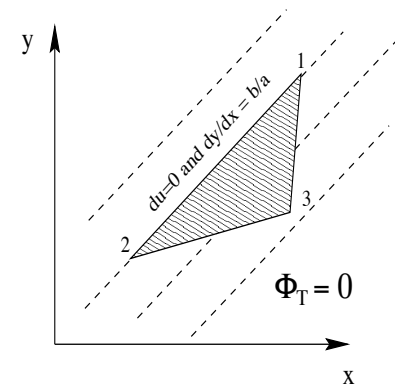
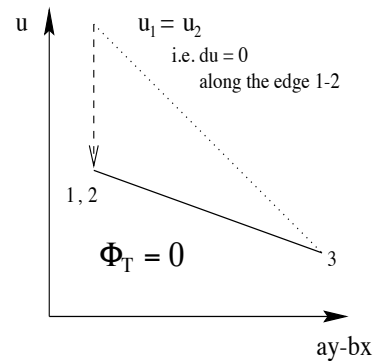
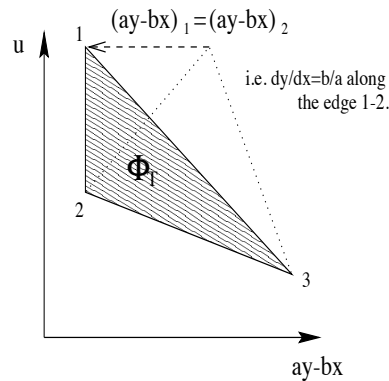
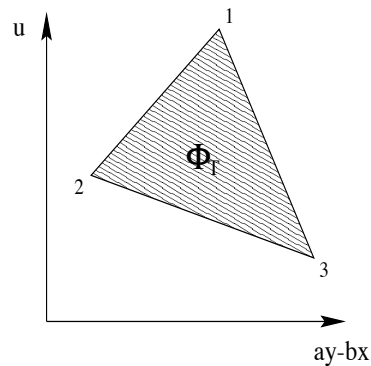
$$u = \text{const.} \quad \text{along} \quad dy/dx = b/a$$



# Fluctuation

$$\Phi_T = \int_T [a \partial_x u + b \partial_y u] dx dy = \frac{1}{2} \sum_{i \in j_T} u_i (a \Delta y_i - b \Delta x_i)$$

*The fluctuation vanishes if the characteristic equation is satisfied along one edge, and it is independent of the third node.*



# Least-Squares Minimization

NORM:

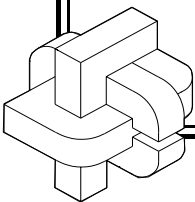
$$\mathcal{F} = \sum F_T = \frac{1}{2} \sum \frac{\Phi_T^2}{S_T}$$

UPDATES:

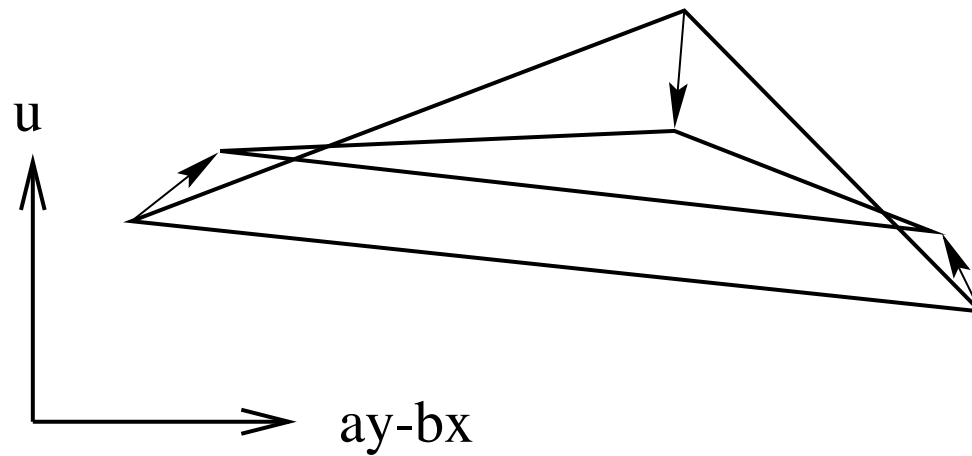
$$\delta u_j = -c_u \frac{\partial \mathcal{F}}{\partial u_j} = \frac{c_u}{2} \sum_{\{T_j\}} (a \Delta y_T - b \Delta x_T) \frac{\Phi_T}{S_T}$$

$$\delta \mathbf{x}_j = -c_x \frac{\partial \mathcal{F}}{\partial \mathbf{x}_j} = \sum_{\{T_j\}} \frac{c_x}{2} \Delta u_T \frac{\Phi_T}{S_T} \begin{bmatrix} -b \\ a \end{bmatrix} + c_x \sum_{\{T_j\}} \frac{F_T}{2S_T} \begin{bmatrix} \Delta y_T \\ -\Delta x_T \end{bmatrix}$$

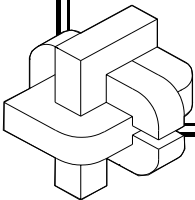
*The node moves in the direction normal to the characteristic.*



# Geometrical View

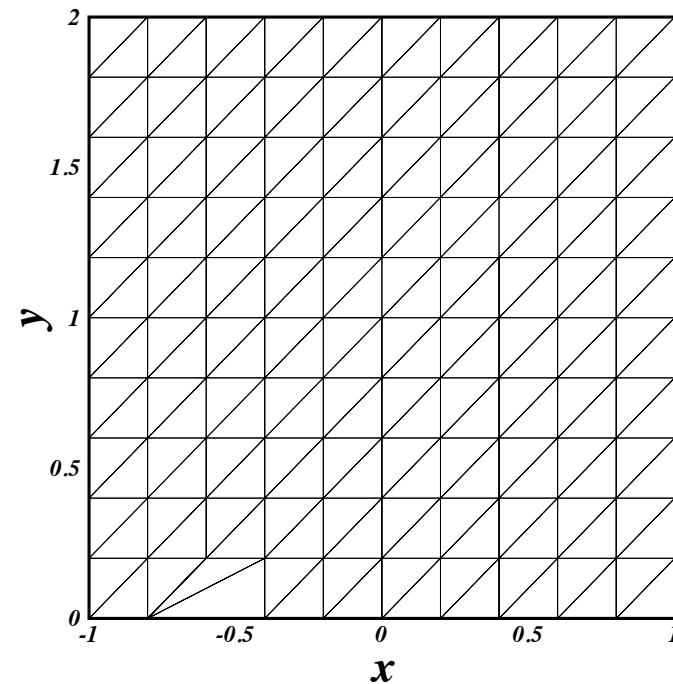
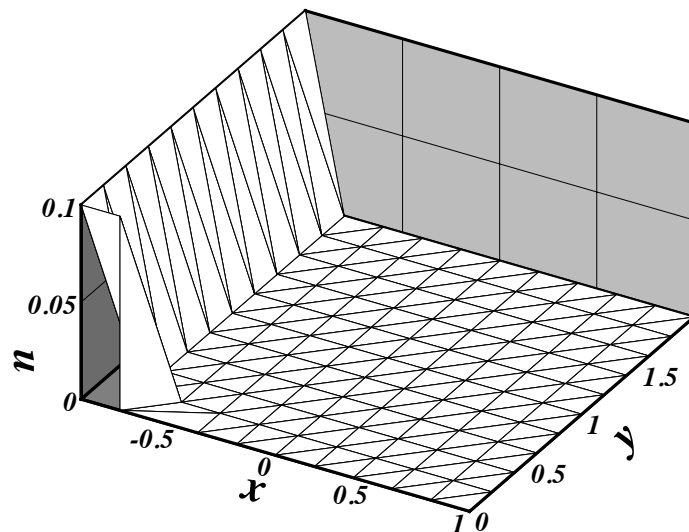


*Minimizing the area of the triangle in the characteristic plane as quickly as possible.*

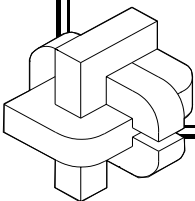




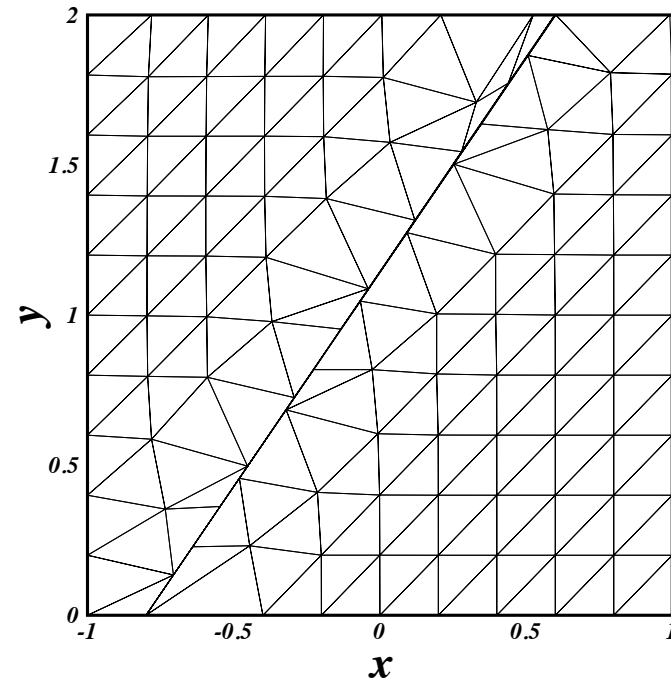
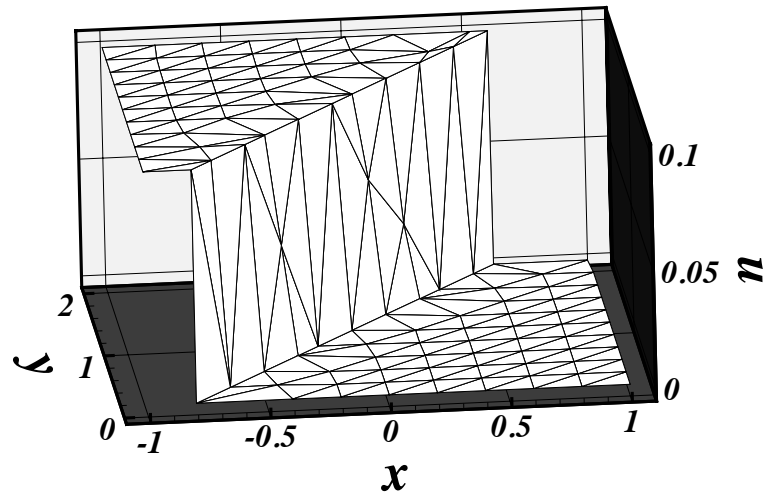
# A Simple Linear Advection



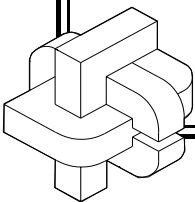
*A degenerate element is introduced to represent a perfect discontinuity.*



# A Simple Linear Advection



*The grid is altered only in an important region.*



# A Linear Hyperbolic System

SMALL PERTURBATION AERODYNAMICS:

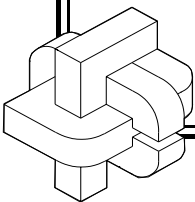
$$(1 - M^2)\partial_x u + \partial_y v = 0, \quad \partial_x v - \partial_y u = 0$$

CHARACTERISTIC EQUATIONS (SUPERSONIC CASE):

$$\beta u + v = \text{const.} \quad \text{along } dy/dx = -1/\beta$$

$$\beta u - v = \text{const.} \quad \text{along } dy/dx = 1/\beta$$

where  $\beta = \sqrt{M^2 - 1}$ .



# Fluctuations

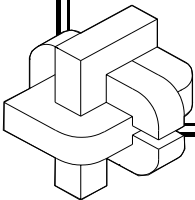
FLUCTUATIONS:

$$\Delta_T = \int_T [\beta^2 \partial_x u - \partial_y v] \, dx \, dy, \quad \Omega_T = \int_T [\partial_x v - \partial_y u] \, dx \, dy$$

CHARACTERISTIC FLUCTUATIONS:

$$C_T = \frac{1}{2} \sum_{i \in j_T} (v + \beta u)_i \Delta(x + \beta y)_i = \Delta_T + \beta \Omega_T$$

$$D_T = \frac{1}{2} \sum_{i \in j_T} (v - \beta u)_i \Delta(x - \beta y)_i = \Delta_T - \beta \Omega_T$$



# The Least-Squares Norm

Minimize the characteristic fluctuations in the least-squares norm.

$$\mathcal{F} = \sum_{T \in \{T\}} F_T = \frac{1}{2} \sum_{T \in \{T\}} \frac{C_T^2 + D_T^2}{S_T}$$

In terms of the original fluctuations.

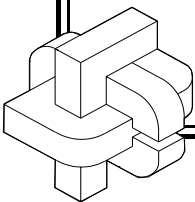
$$\mathcal{F} = \frac{1}{2} \sum_{T \in \{T\}} \frac{\Delta_T^2 + \beta^2 \Omega_T^2}{S_T}$$

In the matrix form,

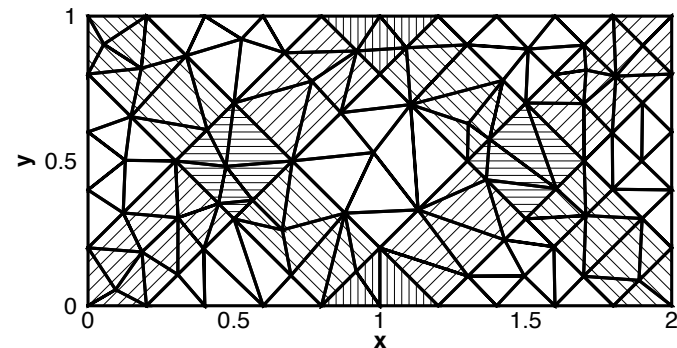
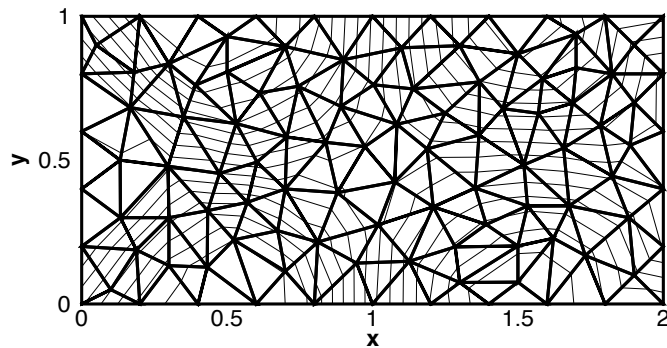
$$\mathcal{F} = \frac{1}{2} \sum \frac{\Phi_T^t Q_T \Phi_T}{S_T}$$

where

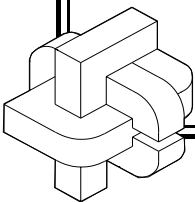
$$\Phi_T = (\Delta_T, \Omega_T)^t, \quad Q_T = \begin{bmatrix} 1 & 0 \\ 0 & \beta^2 \end{bmatrix}$$



# Small Perturbation Aerodynamics

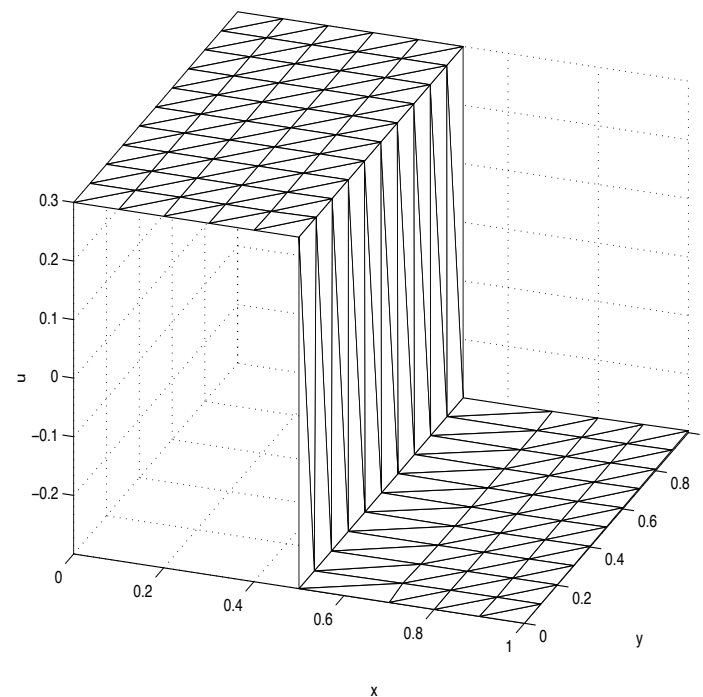
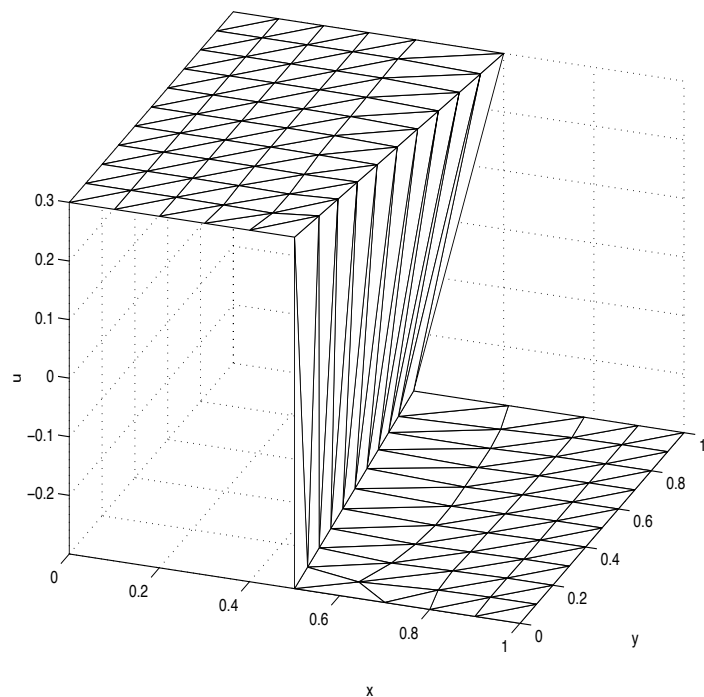


Supersonic flow through a duct,  $M_\infty = \sqrt{2}$  with a small incidence. 18 nodes have been removed on the final grid.

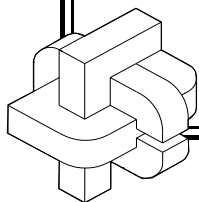


# Burgers' Equation

$$\partial_y u + u \partial_x u = 0$$



$$\Phi_T = \frac{1}{2} \sum_{i \in j_T} u_i (\Delta y_i - \bar{u}_T \Delta x_i) \quad \Phi_T = \frac{1}{2} \sum_{i \in j_T} u_i (\Delta y_i - u_i \Delta x_i)$$



# Quadrature Formulae

CONSERVATION LAWS:

$$\partial_x \mathbf{F}(\mathbf{w}) + \partial_y \mathbf{G}(\mathbf{w}) = 0$$

BILINEAR FLUX FUNCTIONS:

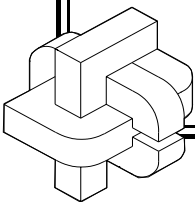
$$\mathbf{F}(\mathbf{w}) = \mathbf{w}^t \mathbf{C} \mathbf{w}, \quad \mathbf{G}(\mathbf{w}) = \mathbf{w}^t \mathbf{D} \mathbf{w}$$

FLUCTUATION:

$$\Phi_{123} = \int \int_{123} [\partial_x \mathbf{F}(\mathbf{w}) + \partial_y \mathbf{G}(\mathbf{w})] dx dy = \oint_{123} [\mathbf{F}(\mathbf{w}) dy + \mathbf{G}(\mathbf{w}) dx]$$

QUADRATURE FORMULAE:

$$\int_1^2 \mathbf{F}(\mathbf{w}) dy = (y_2 - y_1) \left[ \mathbf{w}_1^t \mathbf{C} \mathbf{w}_2 + \frac{\alpha}{2} (\mathbf{w}_1 - \mathbf{w}_2) \mathbf{C} (\mathbf{w}_1 - \mathbf{w}_2) \right]$$





# Formulae for Fluctuation

$$\Phi_{123}(\alpha) = \mathbf{w}_{31}^t(\alpha) \{ \Delta y_3 \mathbf{C} - \Delta x_3 \mathbf{D} \} (\mathbf{w}_3 - \mathbf{w}_1) + \\ \mathbf{w}_{12}^t(\alpha) \{ \Delta y_2 \mathbf{C} - \Delta x_2 \mathbf{D} \} (\mathbf{w}_2 - \mathbf{w}_1)$$

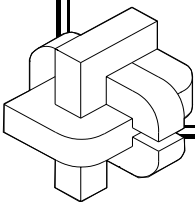
where

$$\mathbf{w}_{ij}(\alpha) = \frac{\alpha}{2}(\mathbf{w}_i + \mathbf{w}_j) + (1 - \alpha)\mathbf{w}_k.$$

Or written

$$\Phi_{123}(\alpha) = \left\{ \frac{\partial \mathbf{F}}{\partial \mathbf{w}} \Delta y_3 - \frac{\partial \mathbf{G}}{\partial \mathbf{w}} \Delta x_3 \right\} (\mathbf{w}_3 - \mathbf{w}_1) + \\ \left\{ \frac{\partial \mathbf{F}}{\partial \mathbf{w}} \Delta y_2 - \frac{\partial \mathbf{G}}{\partial \mathbf{w}} \Delta x_2 \right\} (\mathbf{w}_2 - \mathbf{w}_1)$$

(Note that there are two other rearrangements possible )



# Shock Recognition: $\alpha = 1$

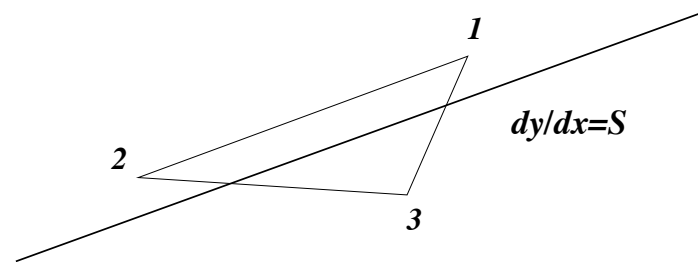
If  $\mathbf{w}_1 = \mathbf{w}_2 = \mathbf{w}_c$ ,

$$\Phi_{123}(\alpha) = \mathbf{w}_{31}^t(\alpha) \{ \Delta y_3 \mathbf{C} - \Delta x_3 \mathbf{D} \} (\mathbf{w}_3 - \mathbf{w}_c).$$

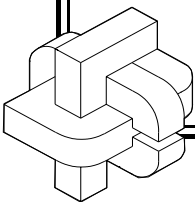
Take  $\alpha = 1 \rightarrow \mathbf{w}_{31}(\alpha) = \bar{\mathbf{w}}_{31} = (\mathbf{w}_3 + \mathbf{w}_c)/2$ .

This corresponds precisely with the Hugoniot condition,

$$\bar{\mathbf{w}}^t (S\mathbf{C} - \mathbf{D}) \Delta \mathbf{w} = 0$$



$\Phi_{123}(1)$  vanishes if the Hugoniot condition is satisfied across a shock and “one edge is aligned with the shock”.



## Entropy-Satisfying Solution: $\alpha = 0$

For Burgers' equation  $\partial_t u + \partial_x f = 0$  where  $f = u^2/2$ ,

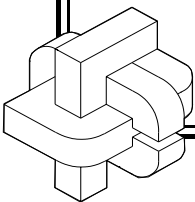
$$\frac{\partial u_j}{\partial t} + \frac{f_{j+1} - f_{j-1}}{2\Delta x} = (1 - \alpha)\Delta x^2 \frac{u_{j+1} - u_{j-1}}{2\Delta x} \frac{u_{j+1} - 2u_j + u_{j-1}}{\Delta x^2}$$

which is a second order discretisation of

$$\frac{\partial u}{\partial t} + \frac{\partial f}{\partial x} = (1 - \alpha)\Delta x^2 \frac{\partial u}{\partial x} \frac{\partial^2 u}{\partial x^2}.$$

**In the case of expansion,  $\frac{\partial u}{\partial x} > 0$ ,**

**$\alpha < 1$  produces positive dissipation  $\rightarrow$  physical solutions.**

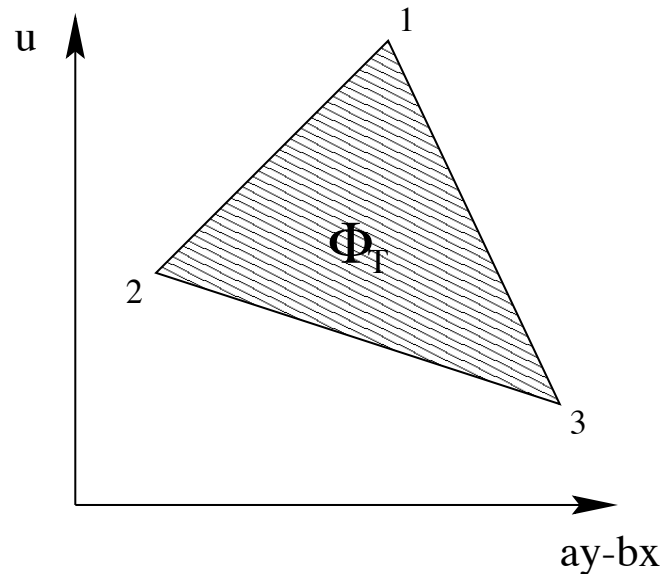


# Characteristic Recognition: $\alpha = 0$

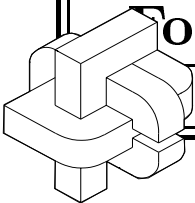
If  $\mathbf{w}_1 = \mathbf{w}_2 = \mathbf{w}_c$ ,

$$\Phi_{123}(\alpha) = \mathbf{w}_{31}^t(\alpha) \{ \Delta y_3 \mathbf{C} - \Delta x_3 \mathbf{D} \} (\mathbf{w}_3 - \mathbf{w}_c).$$

For the special choice  $\alpha = 0 \rightarrow \mathbf{w}_{31}(\alpha) = \mathbf{w}_2$ .



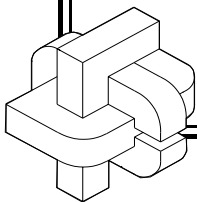
For scalar problems,  $\Phi_{123}(0)$  vanishes if the characteristic



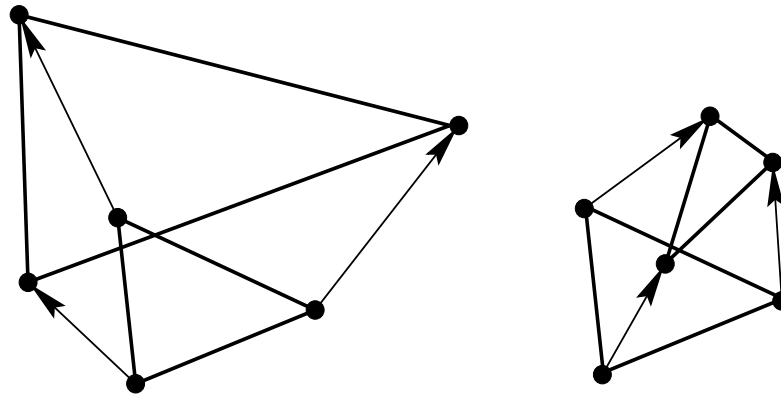
0.5pt=0.482150.5pt

equation  $dw = 0$  is exactly satisfied along one edge

and “that edge is aligned with the exact characteristic”.



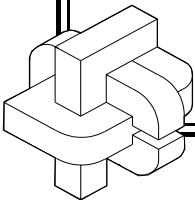
# Detecting Compression/Expansion



Left: Element in expansion. Right: Element in compression.  
Arrows indicate the characteristic speed vecors.

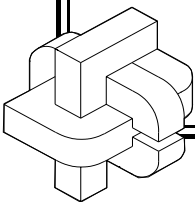
USEFUL QUANTITY:

$$\frac{dS_T}{dt} = \frac{1}{2} \sum_{i=1,2,3} \lambda_i \cdot \mathbf{n}_i$$

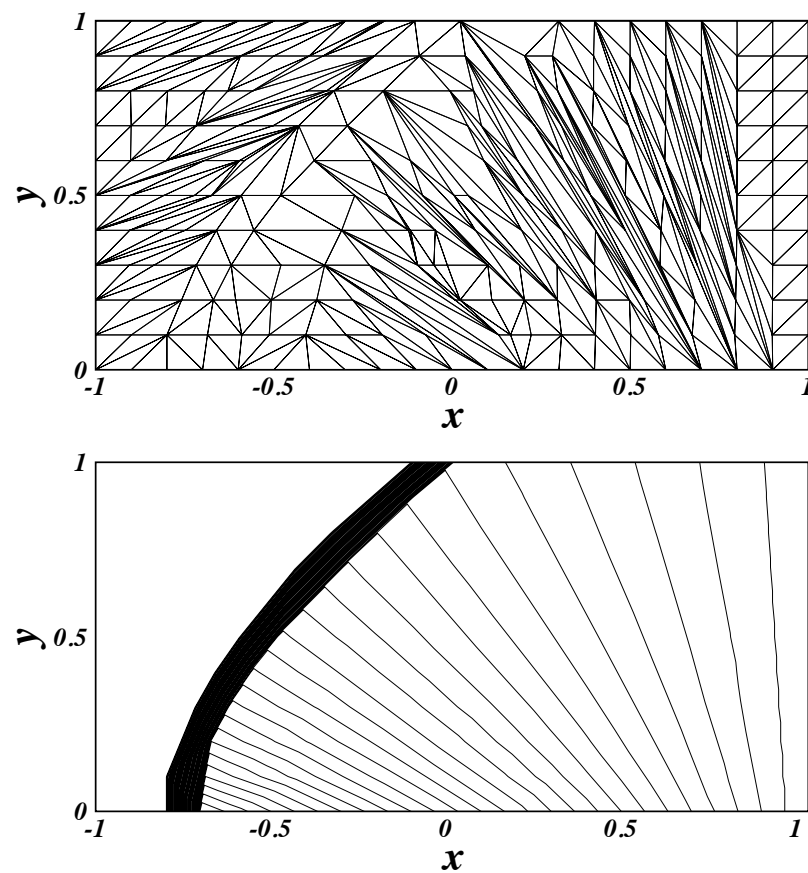


# Solution Procedure

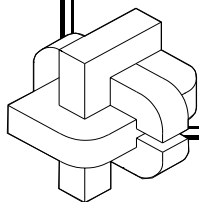
1. Converge the solution on a fixed grid
2. Assign  $\alpha$  to each element, depending on the sign of  $dS_T/dt$
3. Remove undesirable nodes
4. Update solutions and coordinates, with edge swappings interleaved.



# A Curved Shock

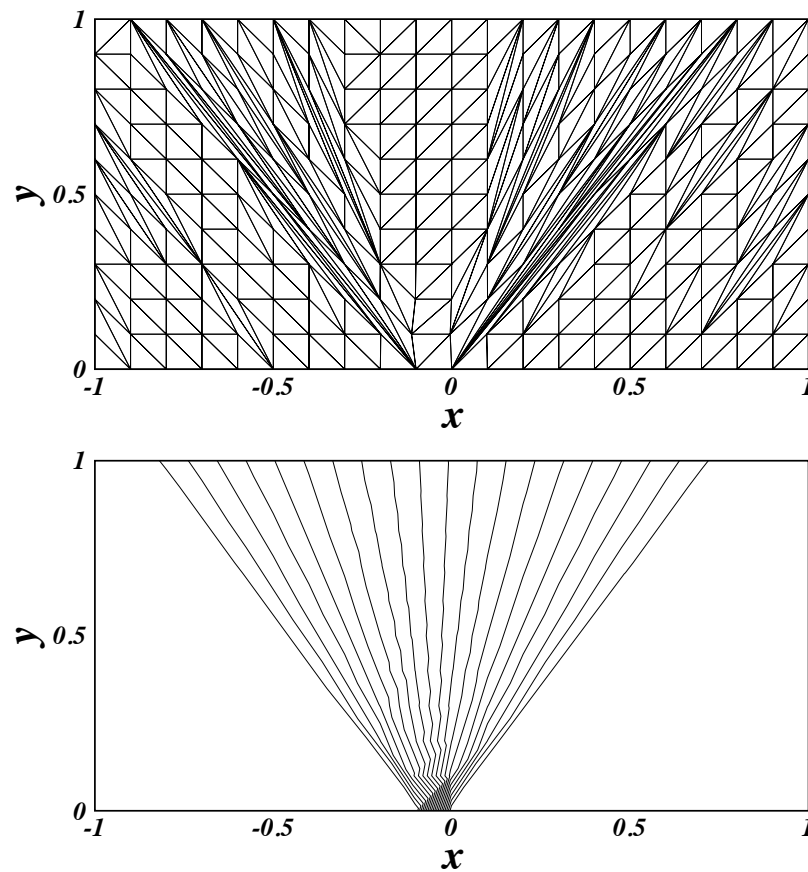


Grid and Solution for a curved shock problem

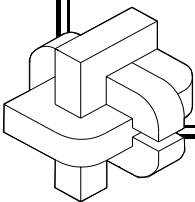




# A Grid-Aligned Rarefaction



Grid and Solution for a rarefaction



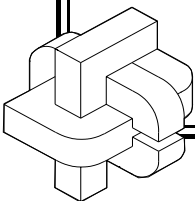
# Cauchy-Riemann System

$$\partial_x \phi - \partial_y \psi = 0$$

$$\partial_y \phi + \partial_x \psi = 0$$

.

Two-dimensional incompressible potential flows with velocity potential  $\phi$ , stream function  $\psi$ . But all 2D elliptic systems are isomorphic with the Cauchy-Riemann system.



# Least-Squares Minimization

FLUCTUATION:

$$U_T = \int_T [\partial_x \phi - \partial_y \psi] \, dx \, dy = \int_T [\partial_\psi y - \partial_\phi x] \, d\phi \, d\psi$$

$$V_T = \int_T [\partial_y \phi + \partial_x \psi] \, dx \, dy = \int_T [-\partial_\phi y - \partial_\psi x] \, d\phi \, d\psi$$

NORM:

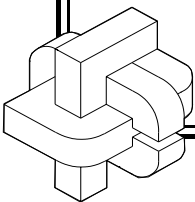
$$\mathcal{F} = \sum_{T \in \{T\}} F_T = \frac{1}{2} \sum_{T \in \{T\}} \frac{U_T^2 + V_T^2}{S_T}$$

Equivalent to minimizing

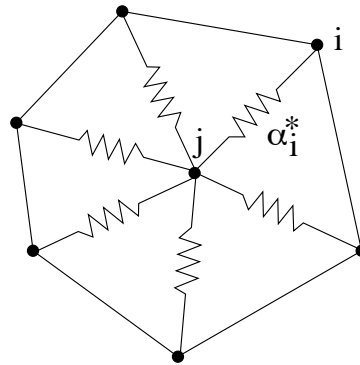
$$\mathcal{F} = \frac{1}{2} \sum_{T \in \{T\}} \iint_T \nabla \phi \cdot \nabla \phi \, dx \, dy + \frac{1}{2} \sum_{T \in \{T\}} \iint_T \nabla \psi \cdot \nabla \psi \, dx \, dy - \sum_{T \in \{T\}} S'_T$$

Equivalent to the standard finite element method applied to

$$\phi_{xx} + \phi_{yy} = 0, \quad \psi_{xx} + \psi_{yy} = 0 \text{ with } \psi = g(x, y), \quad \partial_n \phi = \partial_s \psi$$



# Mesh Movement (Spring Analogy)



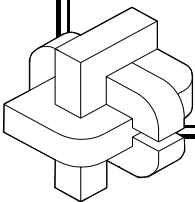
The equations we need to solve,  $\frac{\partial \mathcal{F}}{\partial \mathbf{x}_j} = 0$ , can be written in the iterative form.

$$\mathbf{x}_j^{n+1} = \mathbf{x}_j^n + \frac{\sum_{i \in i_j} \alpha_i^* (\mathbf{x}_i^n - \mathbf{x}_j^n)}{\sum_{i \in i_j} \alpha_i^*} + \frac{1}{2} \frac{\sum_{T \in \{T_j\}} F_T / S_T}{\sum_{i \in i_j} \alpha_i^*} \mathbf{n}_T$$

where

$$\alpha_i^* = \frac{1}{4} (|J_T| \cot \theta_T + |J_{T+1}| \cot \theta_{T+1})$$

$$J_T = (\partial_x \phi \partial_y \psi - \partial_y \phi \partial_x \psi)|_T$$



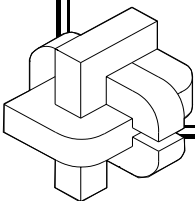
# Cauchy-Riemann System

$$\partial_x u + \partial_y v = 0$$

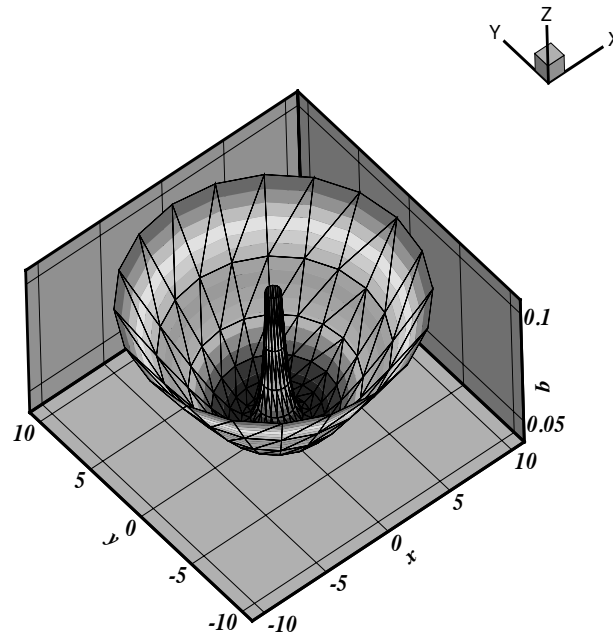
$$\partial_x v - \partial_y u = 0$$

.

“Solutions are continuous in doubly-connected regions.”

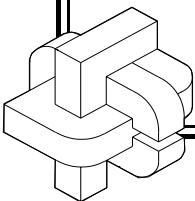


# A Potential Vortex: A Test Problem



The exact solution is  $q = 1/r$ . But the Laplace equations allow another solution.

$$q = 1/r + r$$



# Recovering the Second-Order Accuracy

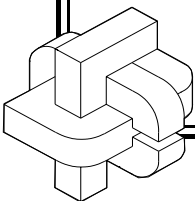
Minimize the “unweighted norm”.

$$\mathcal{F} = \sum_{T \in \{T\}} F_T = \frac{1}{2} \sum_{T \in \{T\}} [U_T^2 + V_T^2]$$

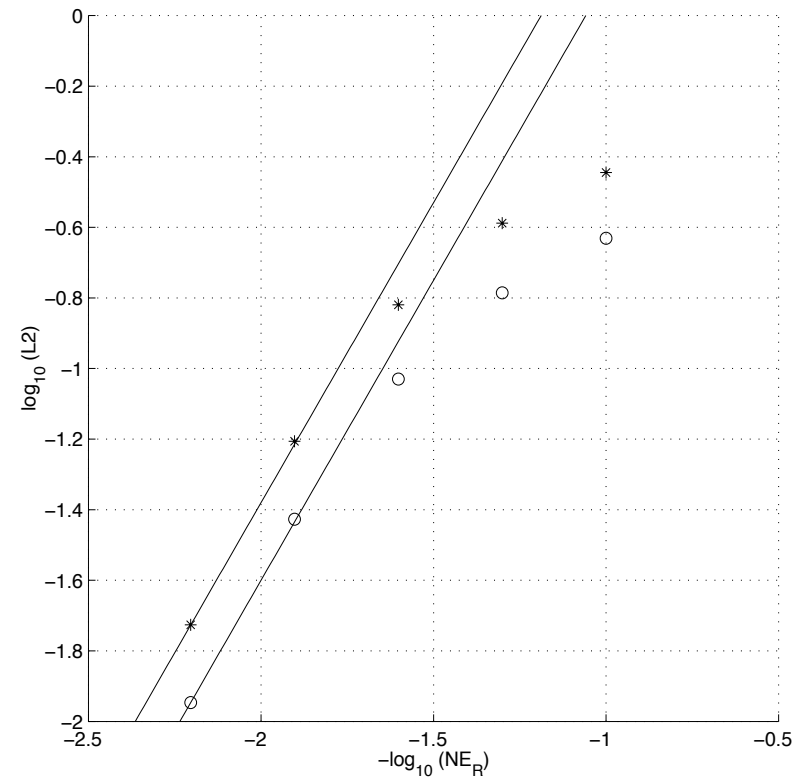
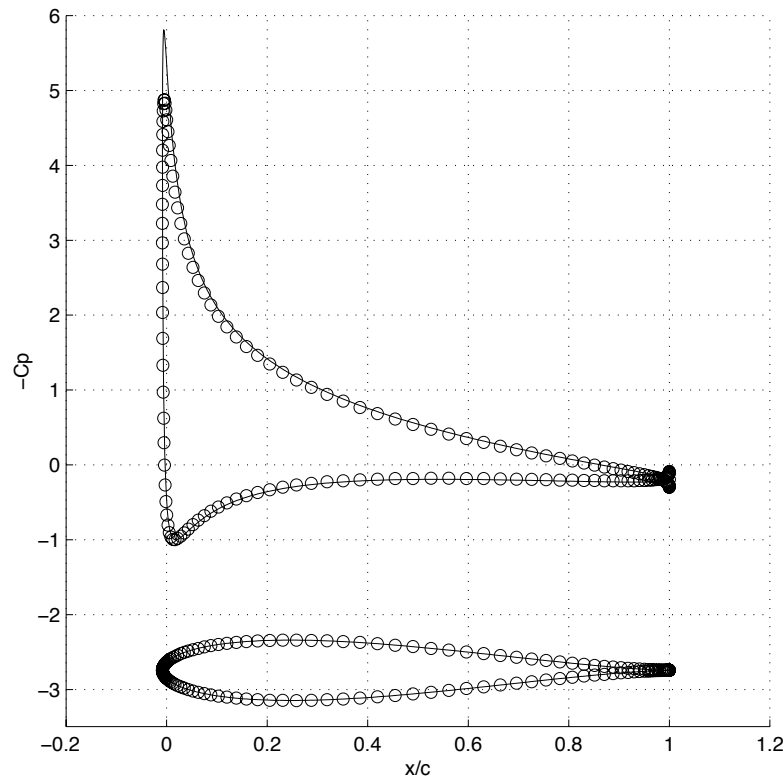
Equivalent to minimizing

$$\mathcal{F} = \frac{1}{2} \sum_{T \in \{T\}} S_T \iint_T \nabla \phi \cdot \nabla \phi \, dx dy + \frac{1}{2} \sum_{T \in \{T\}} S_T \iint_T \nabla \psi \cdot \nabla \psi \, dx dy - \sum_{T \in \{T\}} S_T S'_T$$

“The last term creates the variable-coupling, and it is no longer equivalent to the FEM for the Laplace equations.”

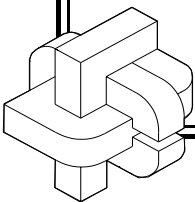


# A Solution by the Unweighted Norm



Solution obtained on an 160x80 O-grid

The convergence rate is 1.7



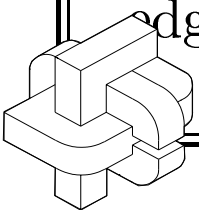


# A Third-Order Least-Squares

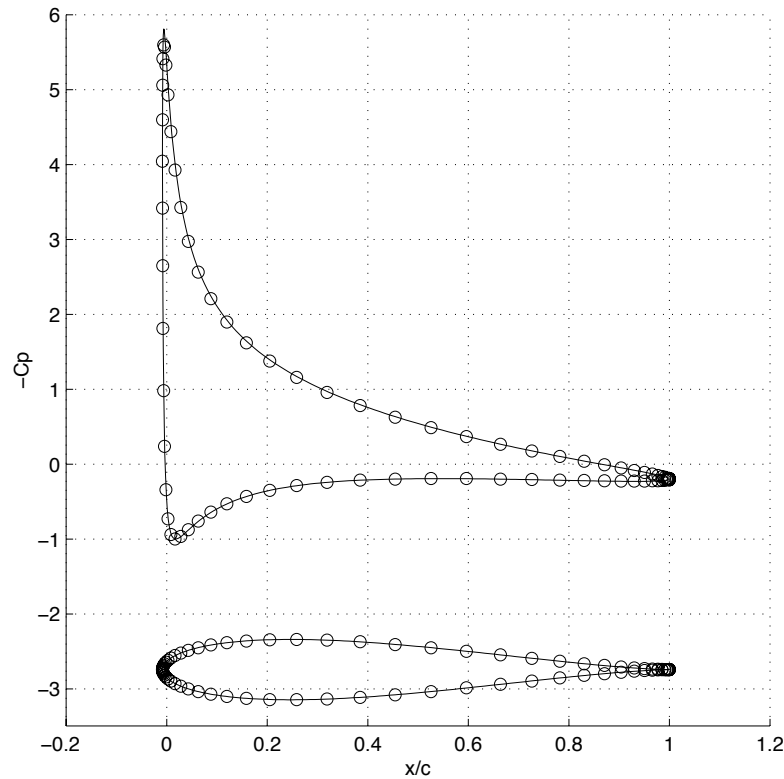
- (1) Evaluate the gradient of each variable.
- (2) Hermite interpolation to get mid point values.
- (3) Simpson's rule to evaluate the fluctuation.

$$\begin{aligned}\Phi_{123} &= \iint_{123} [\partial_x f + \partial_y g] dx dy = \oint_{123} f dy - g dx \\ &= \sum_{edges} (\bar{f} \Delta y - \bar{g} \Delta x) - \frac{1}{12} \sum_{edges} (\Delta p \Delta y - \Delta q \Delta x) .\end{aligned}$$

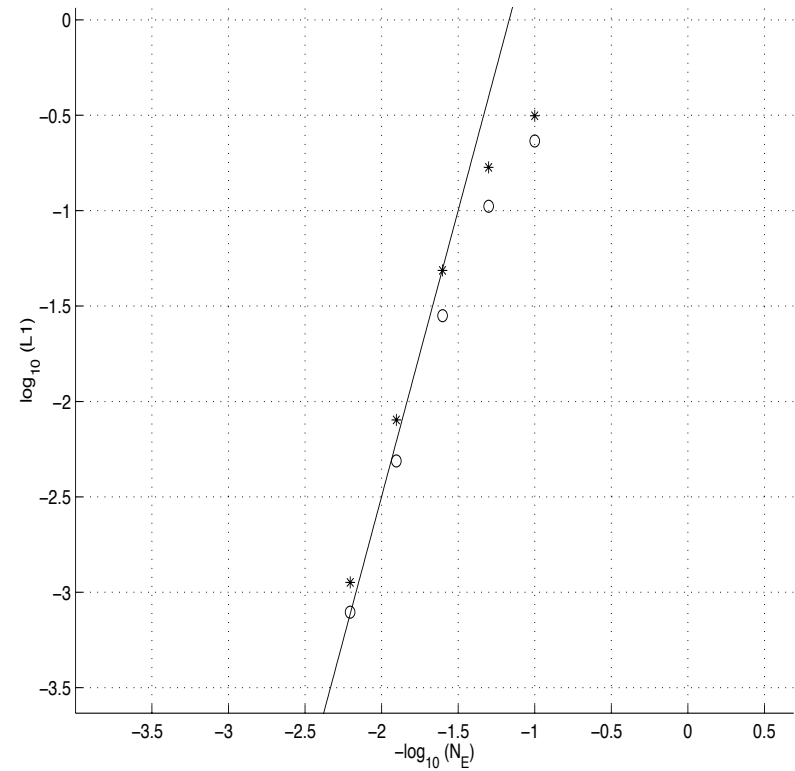
where  $p$  and  $q$  are the directional derivatives of  $f$  and  $g$  along each edge.



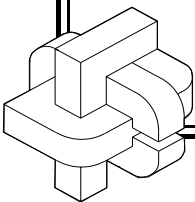
# A Third-Order Least-Squares



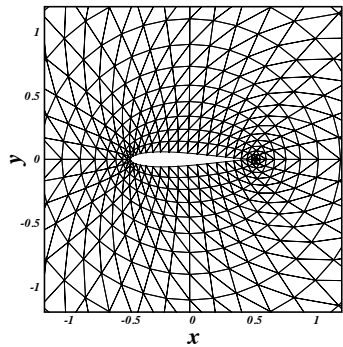
Solution obtained on an 80x40 O-grid



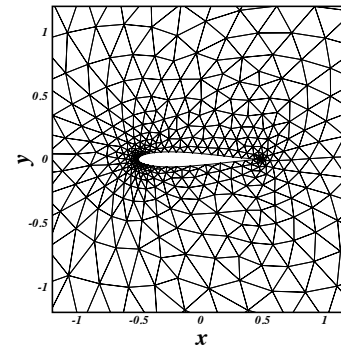
The convergence rate is 3



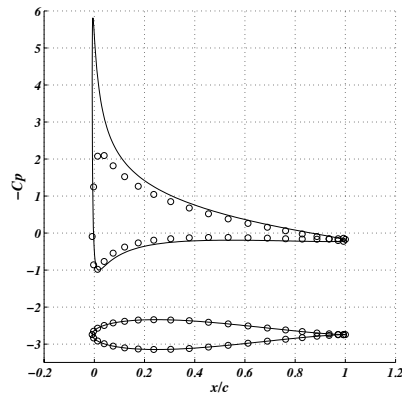
# Moving Mesh Solution



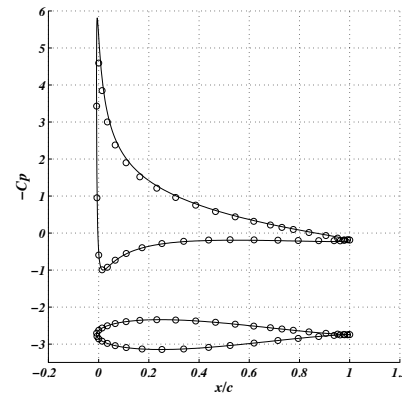
Regular 40x20 Grid



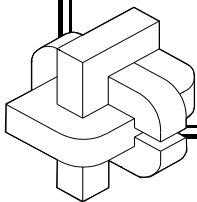
An Adaptive Grid



$C_p$  Distribution



$C_p$  Distribution



# Conclusions and Future Work

- Residual Minimization has the Potential to be a very effective tool for grid adaptation
- Ready for the Euler Equations.
- Finding the Right Norms to Minimize.
- Node Insertion/Removal Procedure.
- Accelerate the Convergence (Multigrid).

