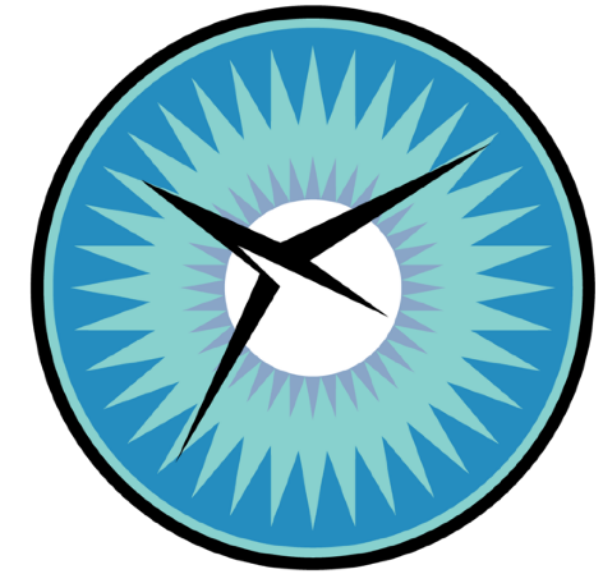


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Updates to Implicit Edge-Based Gradient Methods

Hiroaki Nishikawa, *National Institute of Aerospace*

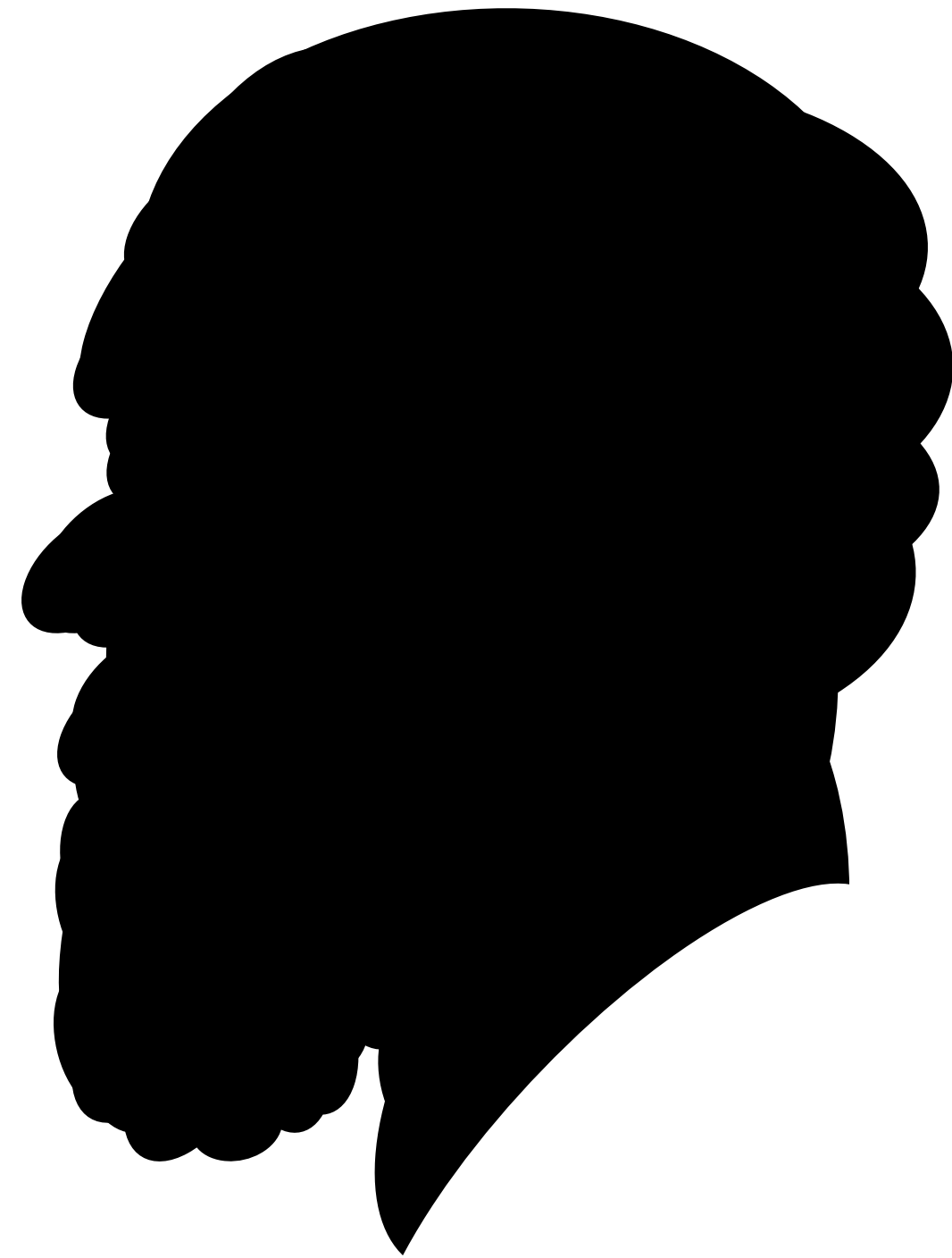
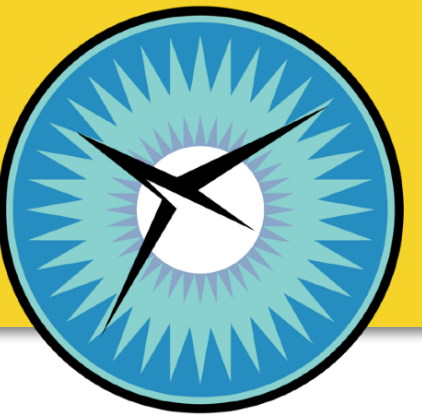
AIAA SciTech, January 10, 2024



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This work was supported partly by the Hypersonic Technology Project, through the Hypersonic Airbreathing Propulsion Branch of the NASA Langley Research Center, under Prime Contract No. 80LARC23DA003



Lesson 1: *Forget that you think you know.*

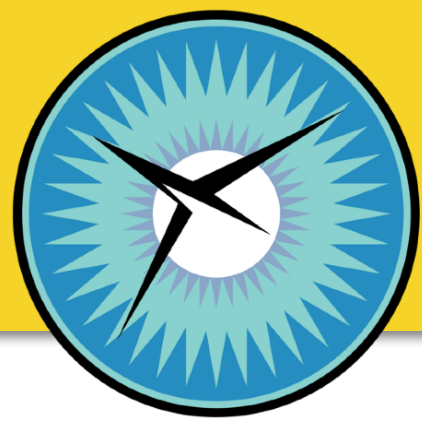
Lesson 2: *See it for yourself.*

Lesson 3: *Understand limits.*

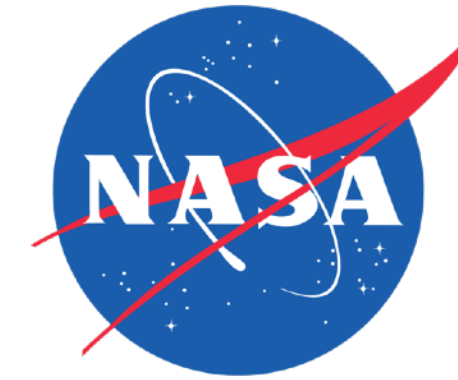
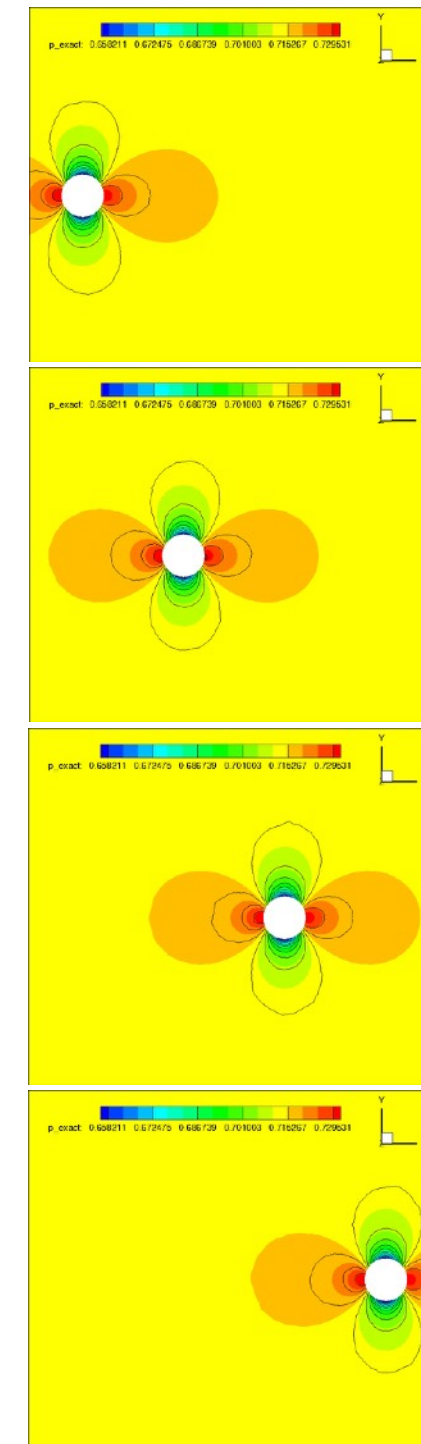
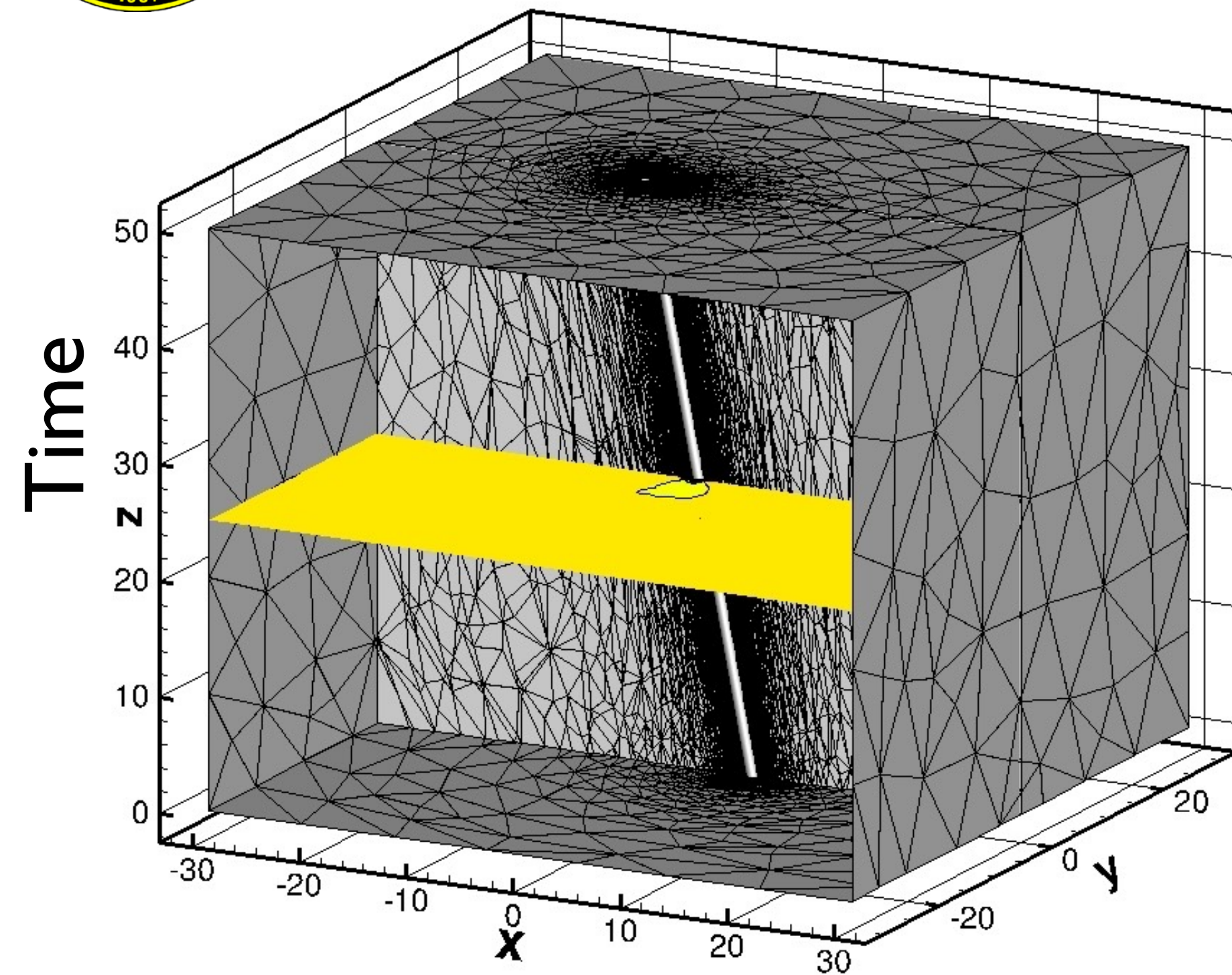
Lesson 1: *Forget that you think you know.*

Economical 3rd-Order Methods

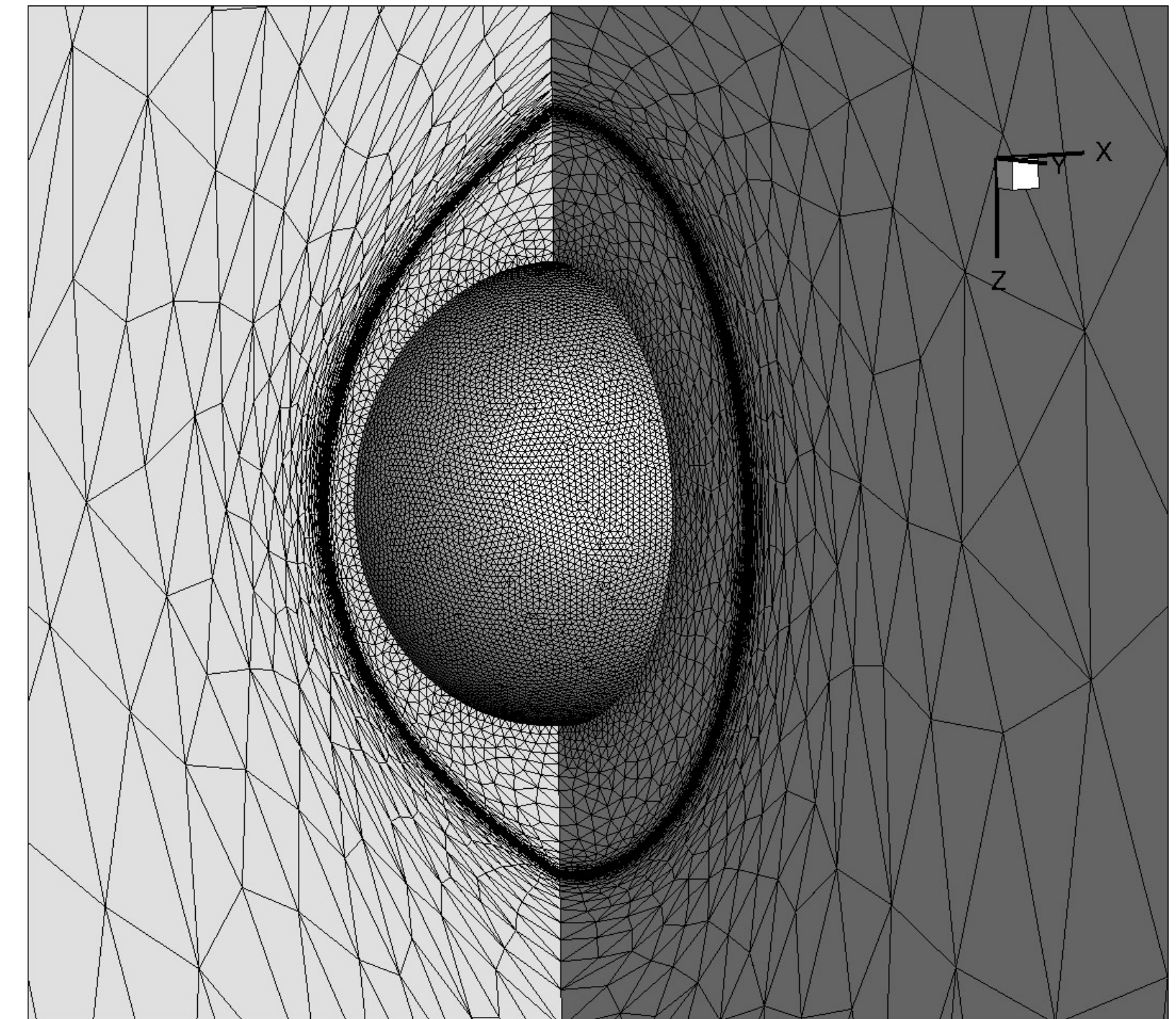
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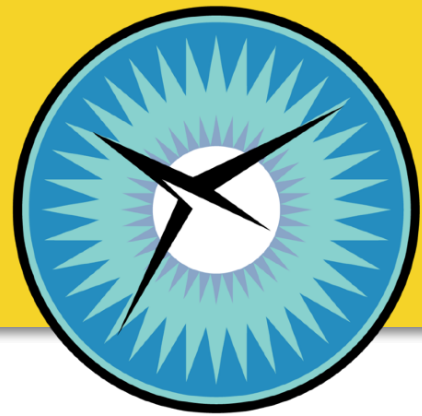
Exploring Space-Time Hyperbolic NS



Improving Unstructured-Grid Methods

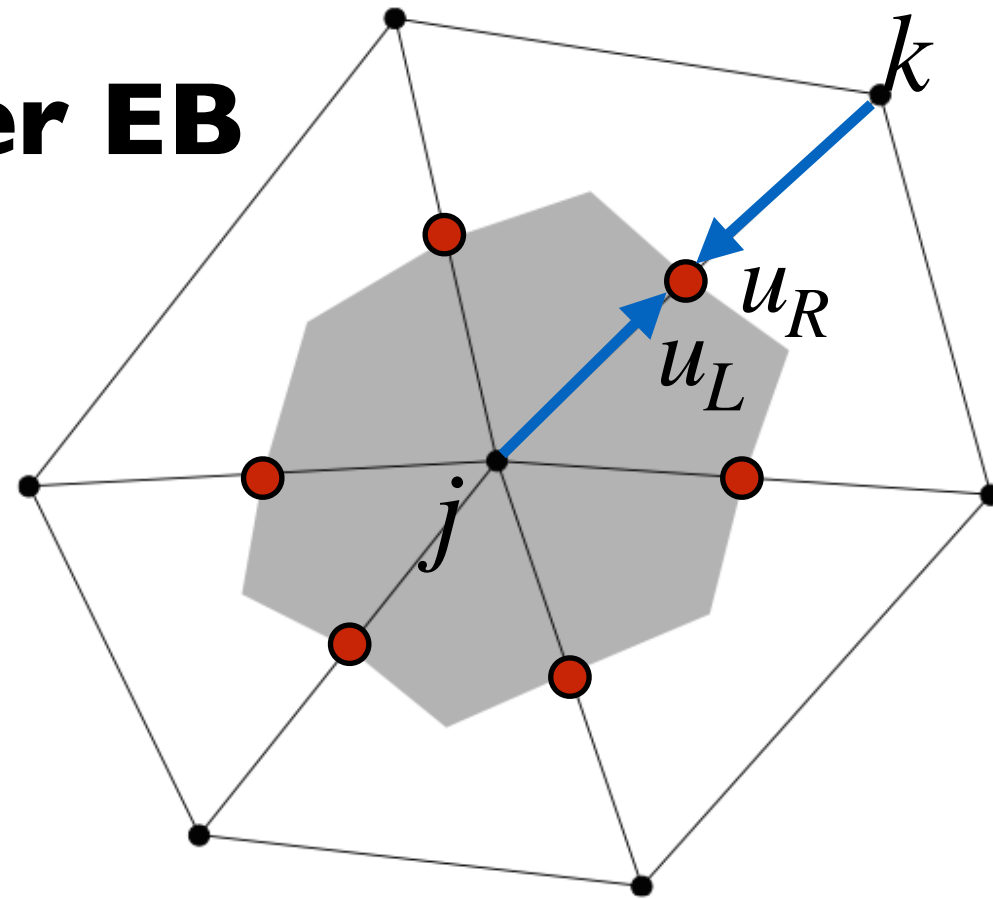


Common interest: *Economical 3rd-order methods* with a single flux per face/edge, not requiring 2nd derivatives at all, towards automated CFD with fully adaptive tetrahedral grids.



Edge-Based Schemes : *A single flux per edge*

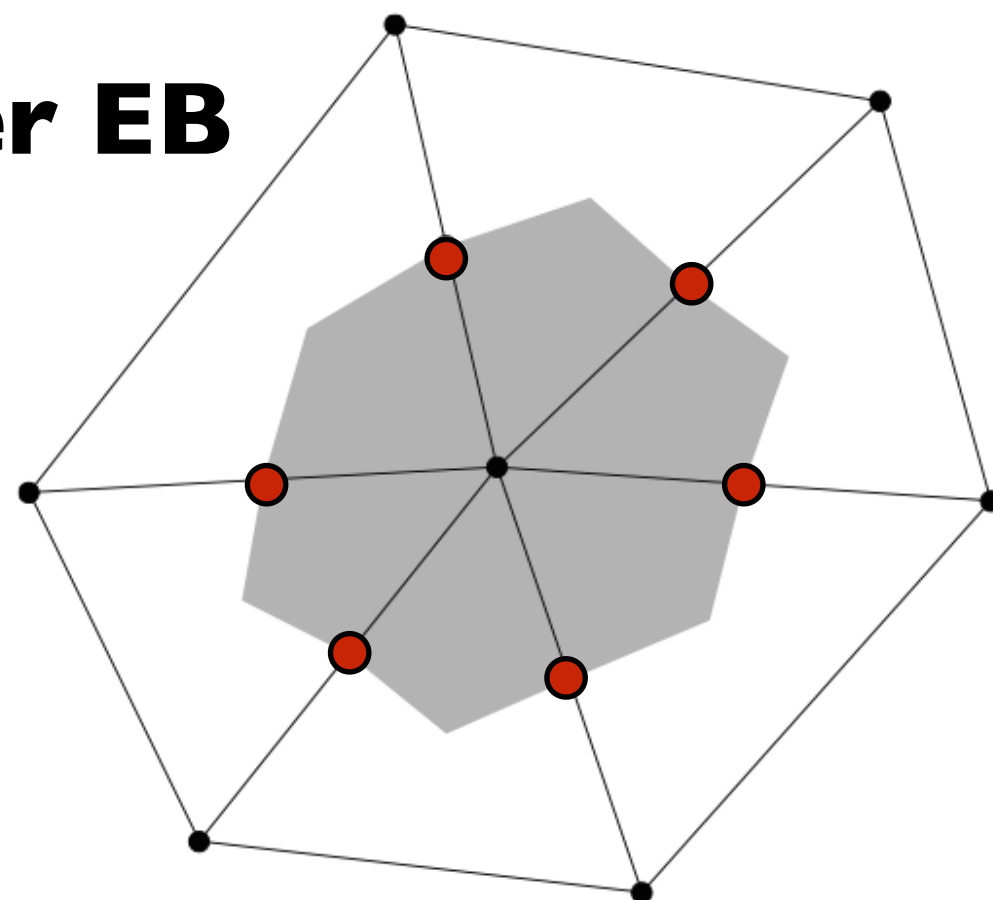
2nd-order EB



$$\frac{1}{V_j} \sum_{k \in \{k_j\}} \phi_{jk}(u_L, u_R) |\mathbf{n}_{jk}| = s(\mathbf{x}_j)$$

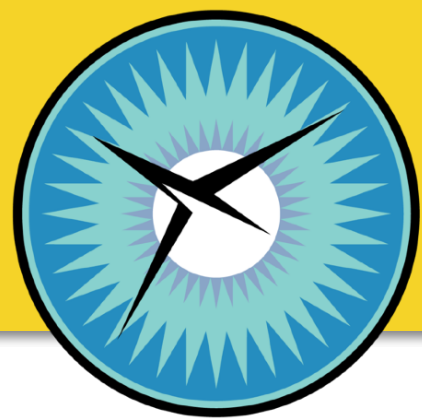
$$u_L = u_j + \frac{1}{2} \mathbf{g}_j \cdot \Delta \mathbf{r}_{jk}, \quad u_R = u_k - \frac{1}{2} \mathbf{g}_k \cdot \Delta \mathbf{r}_{jk}, \quad \Delta \mathbf{r}_{jk} = \mathbf{x}_k - \mathbf{x}_j$$

3rd-order EB



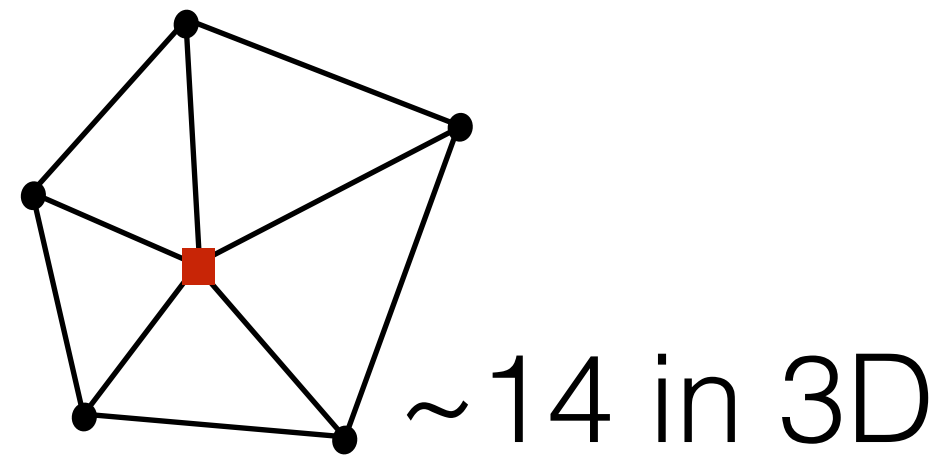
- 3rd-order accurate ([Katz&Sankaran2011](#), [Diskin&Thomas2012](#)) with
- Linear solution/flux reconstruction with **quadratic LSQ gradients**.
 - Accuracy-preserving source quadrature [Nishikawa&Liu JCP2017](#).
- > 2nd derivatives not needed.

for arbitrary triangular/tetrahedral grids.
-> *perfect for adaptive grids.*

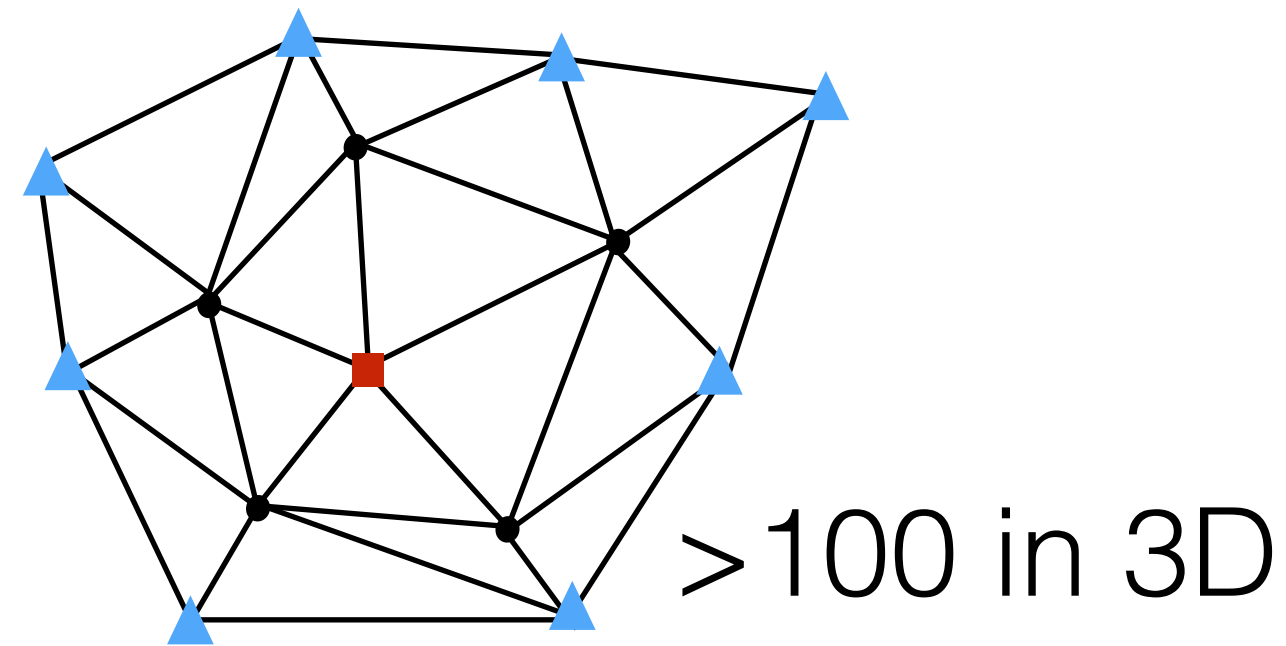


Explicit:

$\mathbf{g}_j^{Linear\ LSQ}$



$\mathbf{g}_j^{Quadratic\ LSQ}$

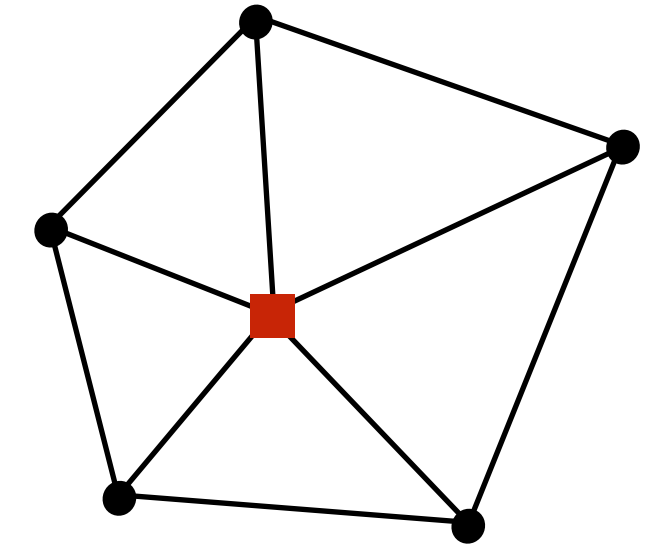


Large stencil: a solver gets more robust but less accurate.

[Haider&Croisille&Courbet NM2009](#)

Implicit:

$$\mathbf{M}_{jj}\mathbf{g}_j + \sum_{k \in \{k_j\}} \mathbf{M}_{jk}\mathbf{g}_k = GG$$



Galerkin: [Löhner\(1994\)](#)

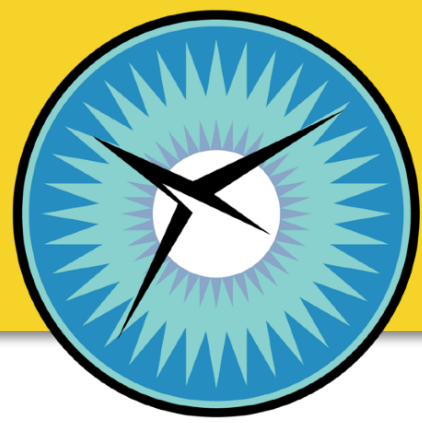
Variational Reconstruction: [Wang et. al., JCP2017](#)

Implicit GG: [Nishikawa, JCP2019](#)

Implicit EB gradient: [Nishikawa, AIAA2020](#)

Advantages of Implicit Gradients

- Flow solver is more stable with a huge gradient stencil.
- Superior gradient accuracy.
- Per-relaxation cost can be lower than LSQ gradients with hundred of neighbors



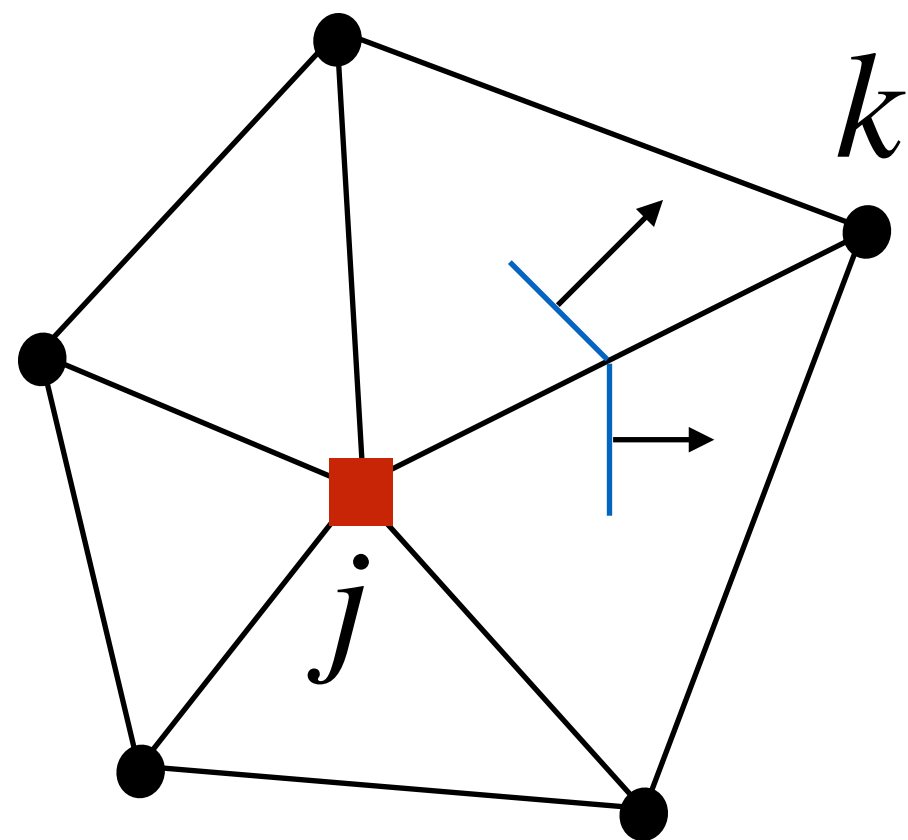
$$\frac{1}{V} \int_V \nabla u dV = \frac{1}{V} \int_V \mathbf{g} dV$$

Accuracy-preserving source quadrature
Nishikawa&Liu JCP2017

$$\frac{1}{V} \oint_{\partial V} u d\mathbf{n} = \frac{1}{V} \int_V \mathbf{g} dV$$

$$\rightarrow \frac{1}{V_j} \sum_{k \in \{k_j\}} \frac{u_L + u_R}{2} \mathbf{n}_{jk} = \frac{1}{V_j} \sum_{k \in \{k_j\}} \left\{ c \mathbf{g}_j + (2 - c) \mathbf{g}'_k \right\} V_{jk}$$

$$\mathbf{g}'_k = \mathbf{g}_k - \nabla \mathbf{g}_j^{LLSQ} (\mathbf{x}_k - \mathbf{x}_j)$$



with

$$u_L = \kappa \frac{u_j + u_k}{2} + (1 - \kappa) [u_j + \mathbf{g}_j \cdot (\mathbf{x}_k - \mathbf{x}_j)]$$

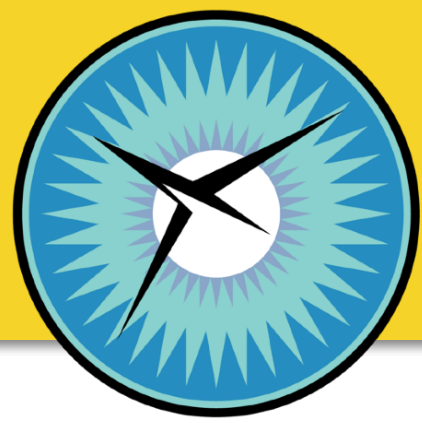
$$u_R = \kappa \frac{u_j + u_k}{2} + (1 - \kappa) [u_k - \mathbf{g}_k \cdot (\mathbf{x}_k - \mathbf{x}_j)]$$

Linear IEBG: Two free parameters: κ and c .

Quadratic IEBG: Remains compact and 3x3 blocks (not 9x9 with 2nd derivatives like others), but no free parameters: $\kappa = 0$ and $c = 13/5$.

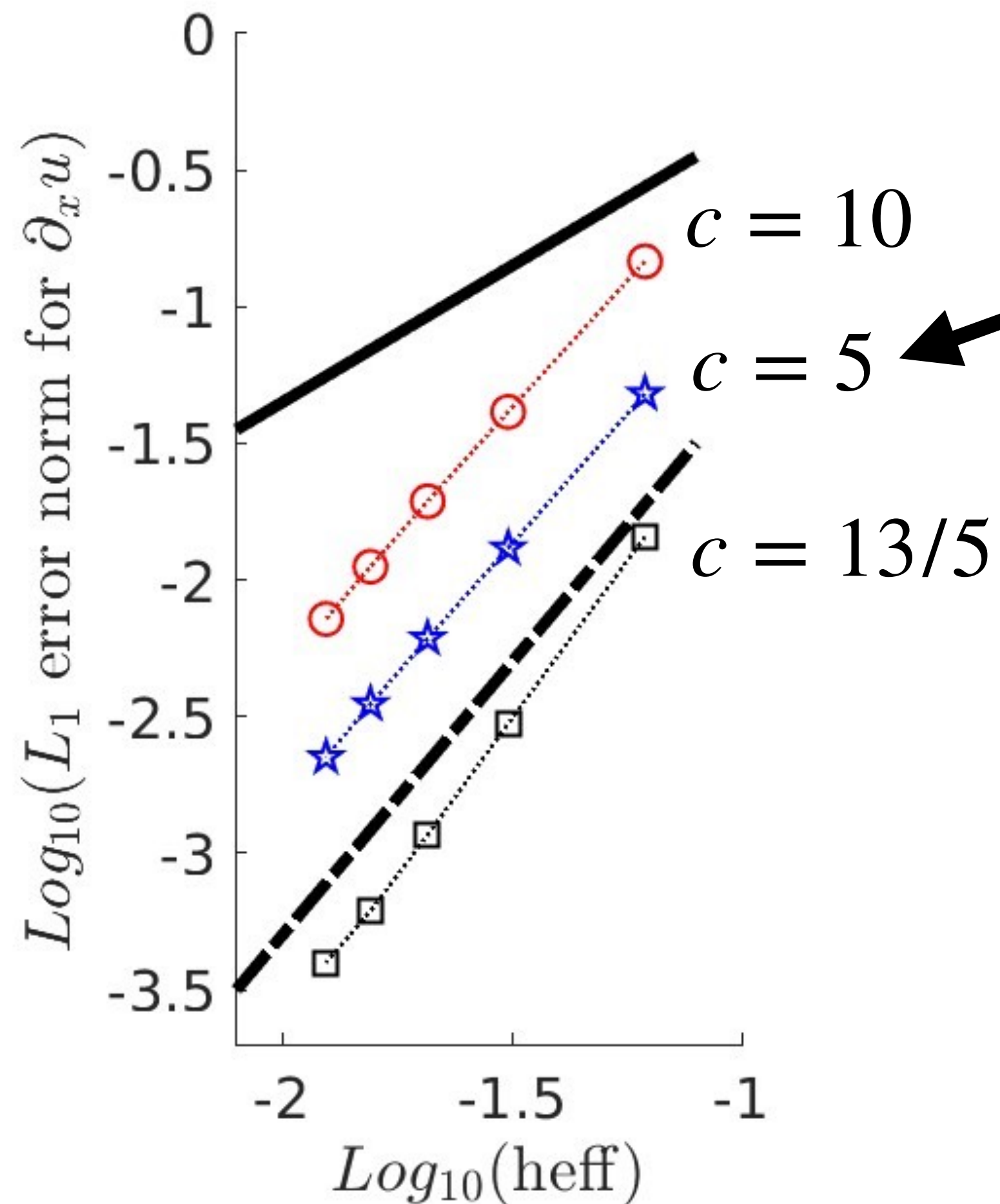
Not very robust... I couldn't adjust anything...

Numerical results show $O(h^2)$ for other values of c



Emmett Padway (NASA) observed 2nd-order gradient accuracy for different values of c (2022).

Later, the author observed the same with his code (2023).



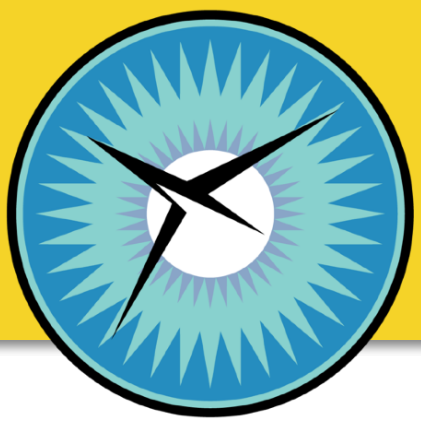
Something is wrong.

Not a bug in a code.

Not a problem in the algorithm.

The problem is that I thought I knew...

Simple if I knew nothing about source quadrature



Forget the source formula and derive:

IEBG system

$$\mathbf{A}_h \mathbf{G}_h = \mathbf{b}$$



Gradient error $\swarrow^{O(1)}$

$$\mathbf{G}_h - \mathbf{G}_{exact} = \mathbf{A}_h^{-1} TE$$

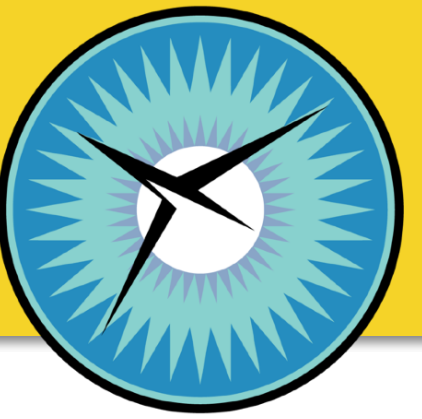
Gradient accuracy order matches TE order.

meaning that QIEBG must have $TE = O(h^2)$. It actually does for *any* c :

$$\frac{1}{V_j} \sum_{k \in \{k_j\}} \frac{u_L + u_R}{2} \mathbf{n}_{jk} - \frac{1}{V_j} \sum_{k \in \{k_j\}} \left\{ c \mathbf{g}_j + (2 - c) \mathbf{g}'_k \right\} V_{jk} = \nabla u - \mathbf{g} + \underline{O(h^2)}$$

*So, we have 2nd-order gradient accuracy for any c .
That's good news: I can try to adjust it for robustness.*

So, what was wrong?

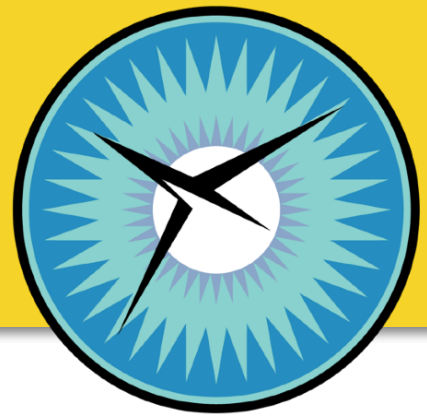


Let me skip this to avoid confusion.

(Please read the paper or ask me.)

Better understanding by forgetting that I think I know.

Lesson 2: *See it for yourself.*



$$\frac{1}{V_j} \sum_{k \in \{k_j\}} \frac{u_L + u_R}{2} \mathbf{n}_{jk} = \frac{1}{V_j} \sum_{k \in \{k_j\}} \left\{ c \mathbf{g}_j + (2 - c) \mathbf{g}'_k \right\} V_{jk}$$

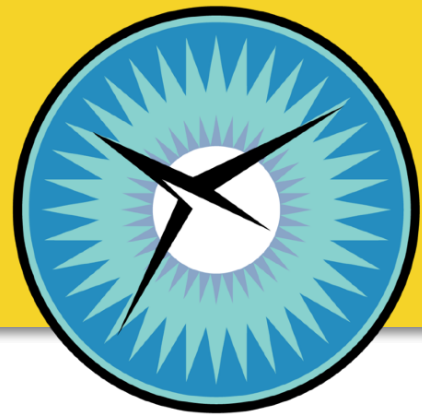
In 2020, I computed $\mathbf{M}_{jj}$ and $\mathbf{M}_{jk}$
by automatic differentiation

$$\longrightarrow \mathbf{M}_{jj} \mathbf{g}_j + \sum_{k \in \{k_j\}} \mathbf{M}_{jk} \mathbf{g}_k = GG$$

Solve the linear system by a relaxation scheme: e.g., Gauss Seidel.

$$\mathbf{g}_j^{n+1} = -\mathbf{M}_{jj}^{-1} \left[\sum_{k \in \{k_j\}} \mathbf{M}_{jk} \mathbf{g}_k^n \right] + \mathbf{M}_{jj}^{-1} GG$$

This relaxation fails for some sets of parameters κ and c . Why? I didn't know...



To gain some insight on how IEBG fails when it fails,
I decided to derive the block matrix $\mathbf{M}_{jj}$ and see what it looks like.

Linear IEBG:

$$\begin{aligned}\mathbf{M}_{jj} &= \frac{\kappa - 1}{4V_j} \sum_{k \in \{k_j\}} (\mathbf{n}_{jk} \otimes \Delta \mathbf{r}_{jk}) + \frac{c}{2V_j} \sum_{k \in \{k_j\}} V_{jk} \mathbf{I} \\ &= \frac{\kappa - 1}{2V_j} V_j \mathbf{I} + \frac{c}{2V_j} V_j \mathbf{I} \\ &= \left(\frac{c + \kappa - 1}{2} \right) \mathbf{I}\end{aligned}$$

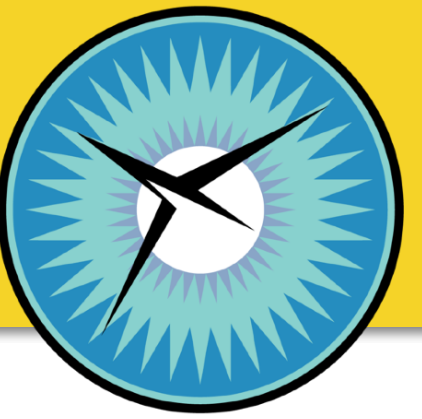
$$\frac{1}{2} \sum_{k \in \{k_j\}} (\mathbf{n}_{jk} \otimes \Delta \mathbf{r}_{jk}) = V_j \mathbf{I}$$

for arbitrary triangular/tetrahedral grids:

Not a matrix but a scalar! Very easy to invert: $\mathbf{M}_{jj}^{-1} = \left(\frac{2}{c + \kappa - 1} \right) \mathbf{I}$ (similarly for QIEBG).

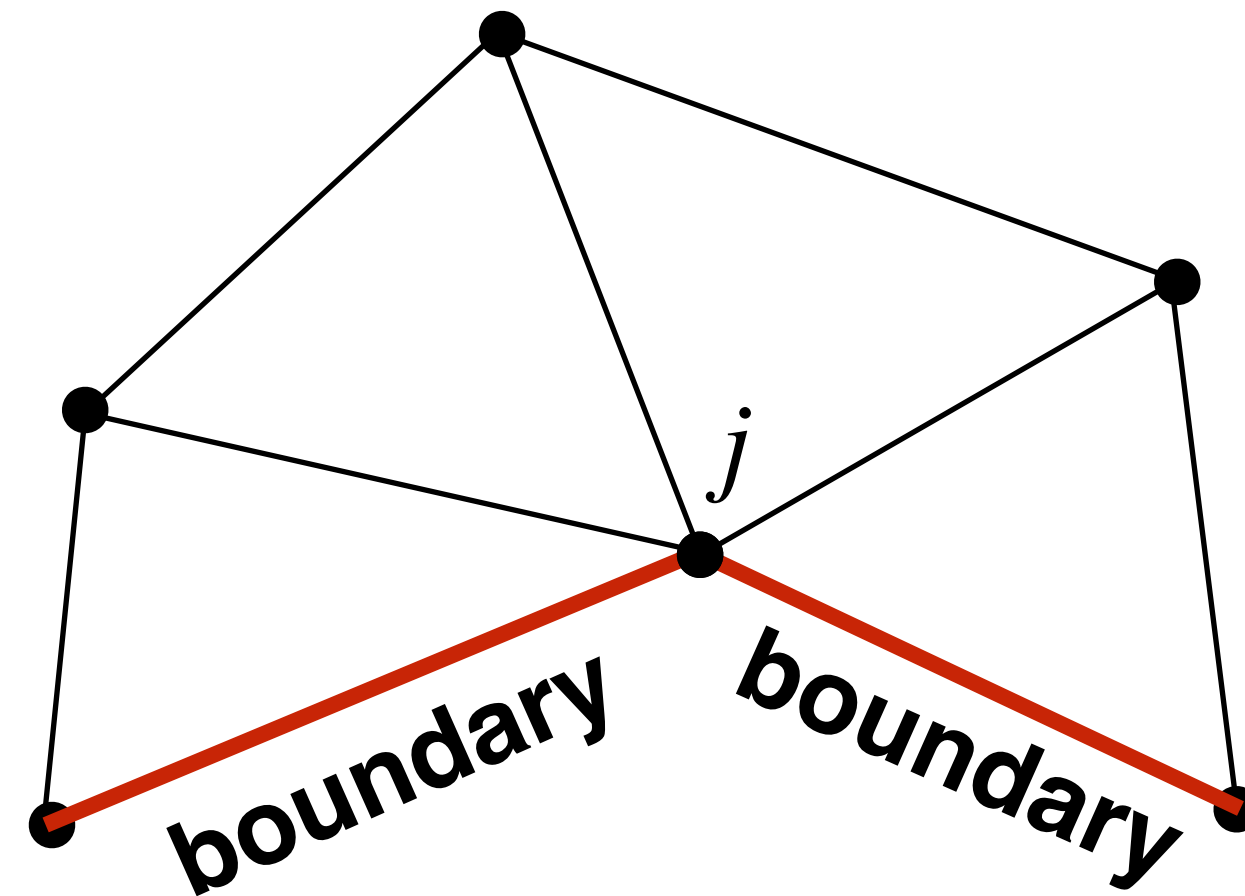
Oh, it fails when $c + \kappa - 1 = 0$!

True also at boundary nodes



This remains true while an edge is collapsed:

$$\mathbf{M}_{jj} = \left(\frac{c + \kappa - 1}{2} \right) \mathbf{I}$$



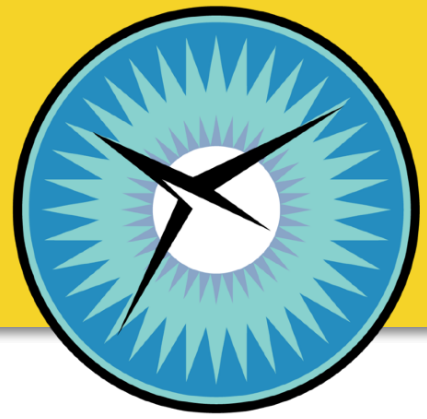
Achieved great simplification by seeing it for myself.

Lesson 3: *Understand limits.*

What values are allowed for κ and c ? I didn't know well...

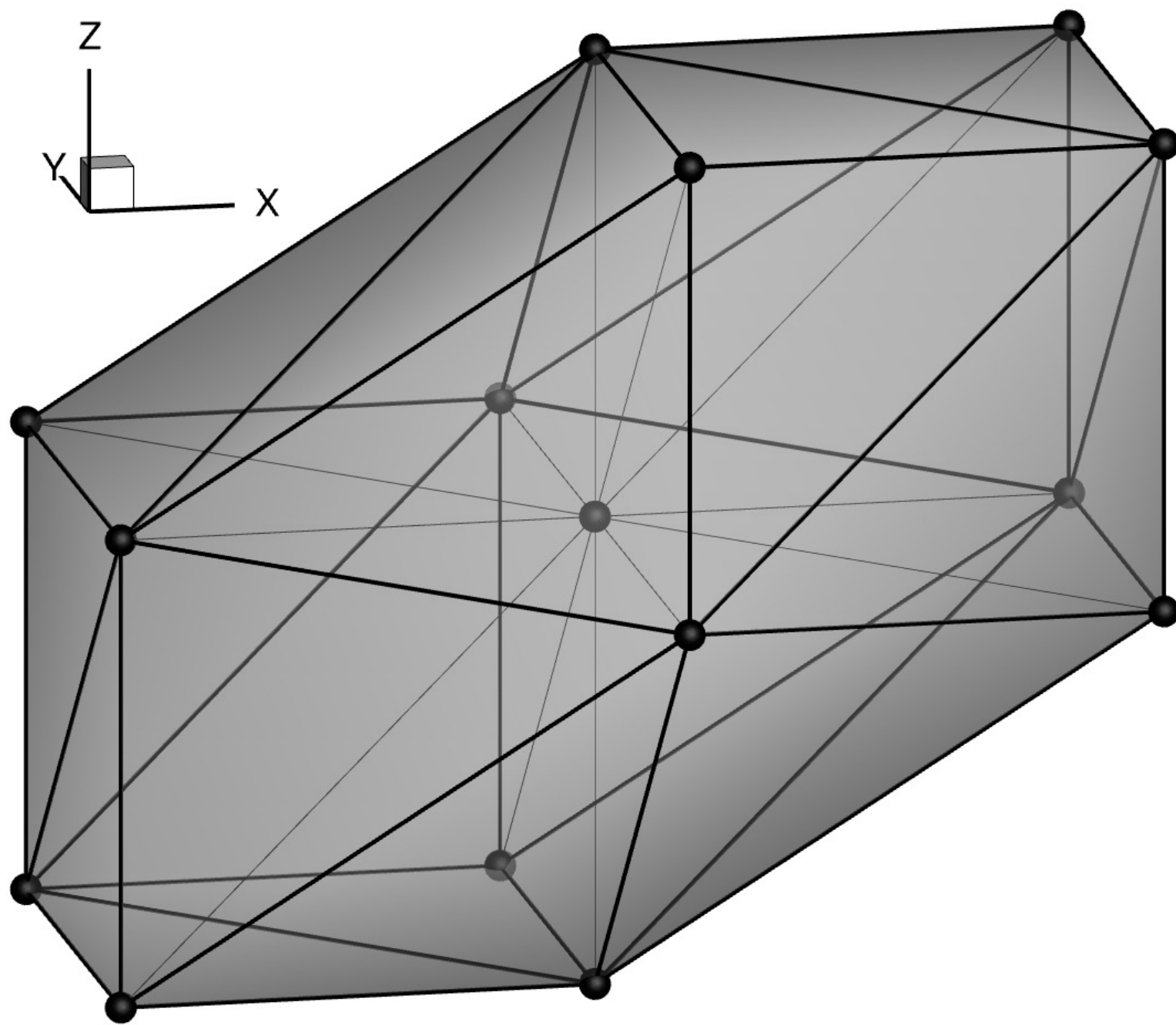
-> Look at accuracy and stability.

Accuracy: Good to find special values



On a regular tetrahedral grid: Linear IEBG = Quadratic IEBG

$$\mathbf{Res}_j = \frac{5}{36} \left(c - \frac{13 - 9\kappa}{5} \right) \left(\partial_{xx} + \partial_{yy} + \partial_{zz} - \partial_{xy} - \partial_{yz} + \partial_{zx} \right) (\nabla u) h^2 + O(h^4).$$



Regular tetrahedral grid

$$c = \frac{13 - 9\kappa}{5}$$

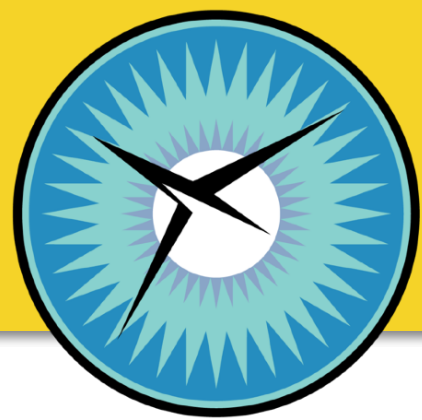
<- 4th-order accurate.

In this work, we set $\kappa = 0$ and $c = \frac{13}{5}$,

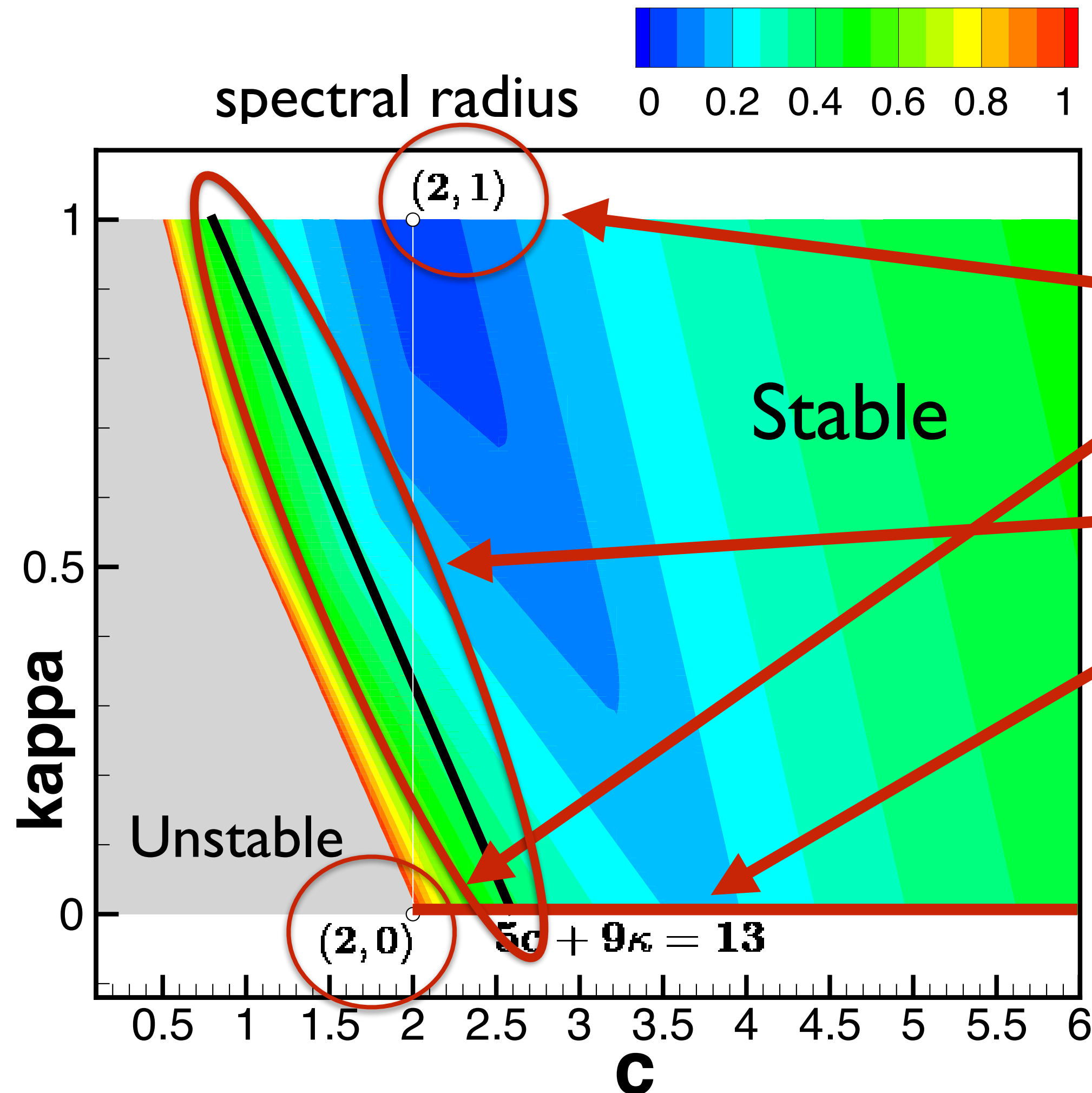
but a larger value $c = 5$ was needed for robustness.

Is it supposed to work?

Stability: Good to understand limits



Performed a Fourier stability analysis for the Gauss-Seidel scheme.



So, we see

- Explicit at $(c, \kappa) = (2, 1)$. Zero spectral radius!
- Linear IEBG is 2nd-order accurate at $(2, 0)$ but unstable.
- 4th-order IEBG is stable: $5c + 9\kappa = 13$
- Quadratic IEBG ($\kappa = 0$) is stable for $c > 2$.

So, we can safely set $c = 5$.

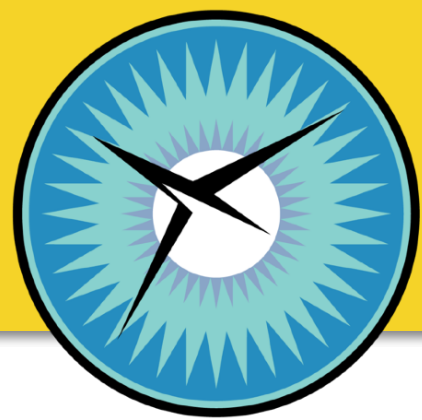
A better guide on how to choose parameters.

Results

Mach 0.3 smooth bump

(Accuracy verification and supersonic/hypersonic problems in the paper.)

Mach 0.3 Subsonic Flow over a smooth bump

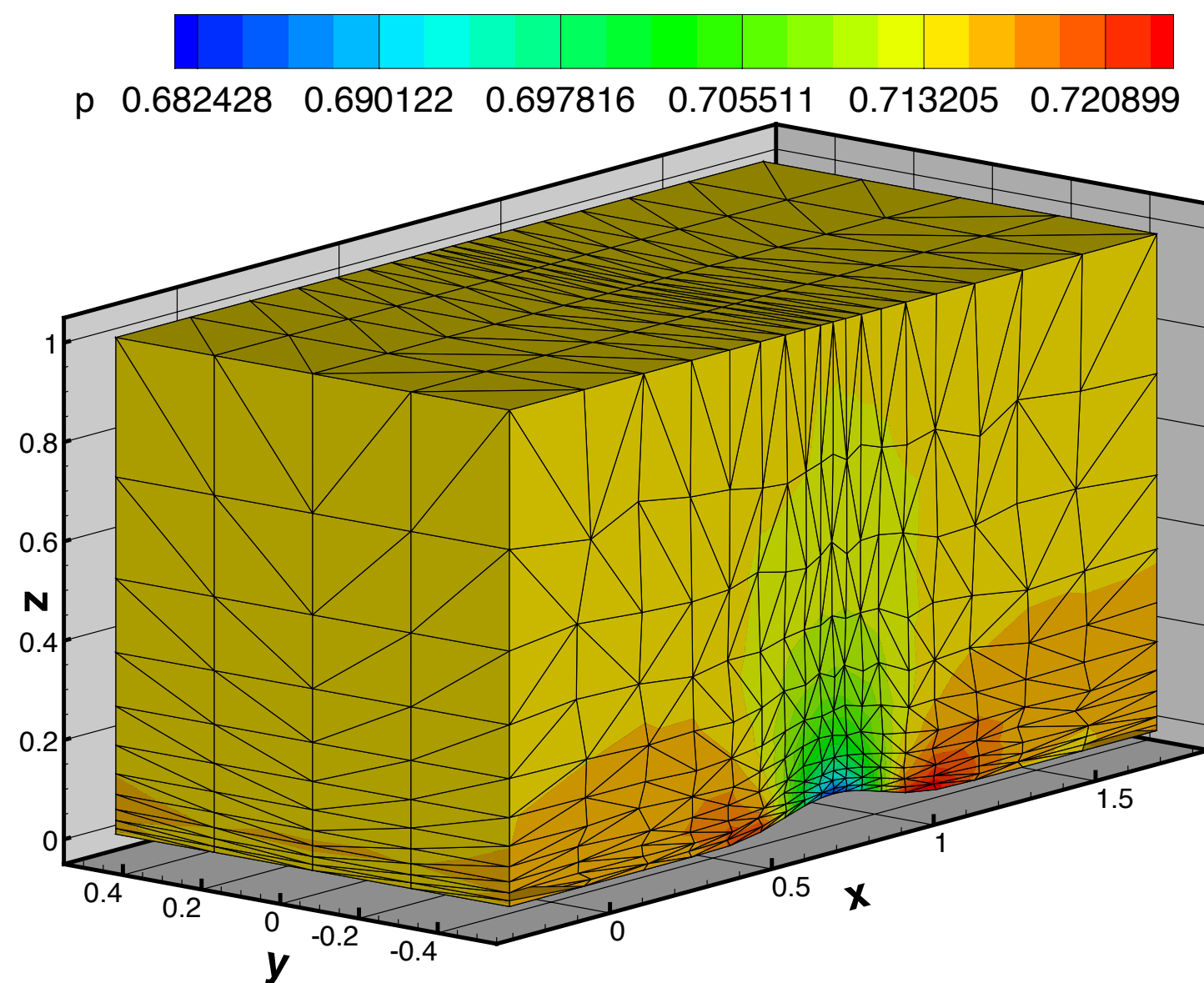


Solve Euler by 2nd/3rd-order EB methods:

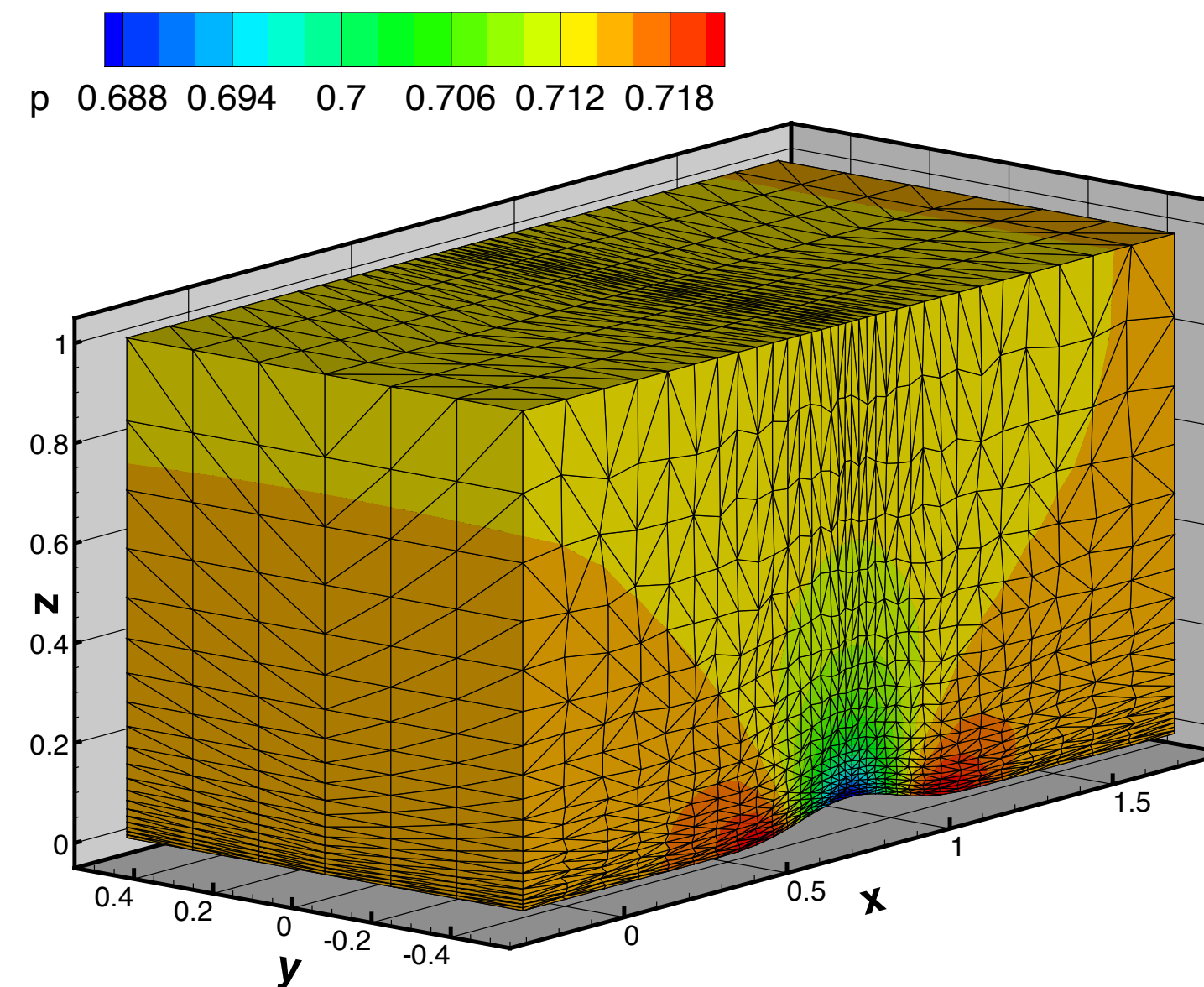
(1) $\mathbf{G}^{n+1} = \mathbf{G}^n + \Delta\mathbf{G}$: Update gradients by one Gauss-Seidel relaxation.

(2) $\mathbf{U}^{n+1} = \mathbf{U}^n + \Delta\mathbf{U}$: Update solutions by one iteration of implicit Euler solver.

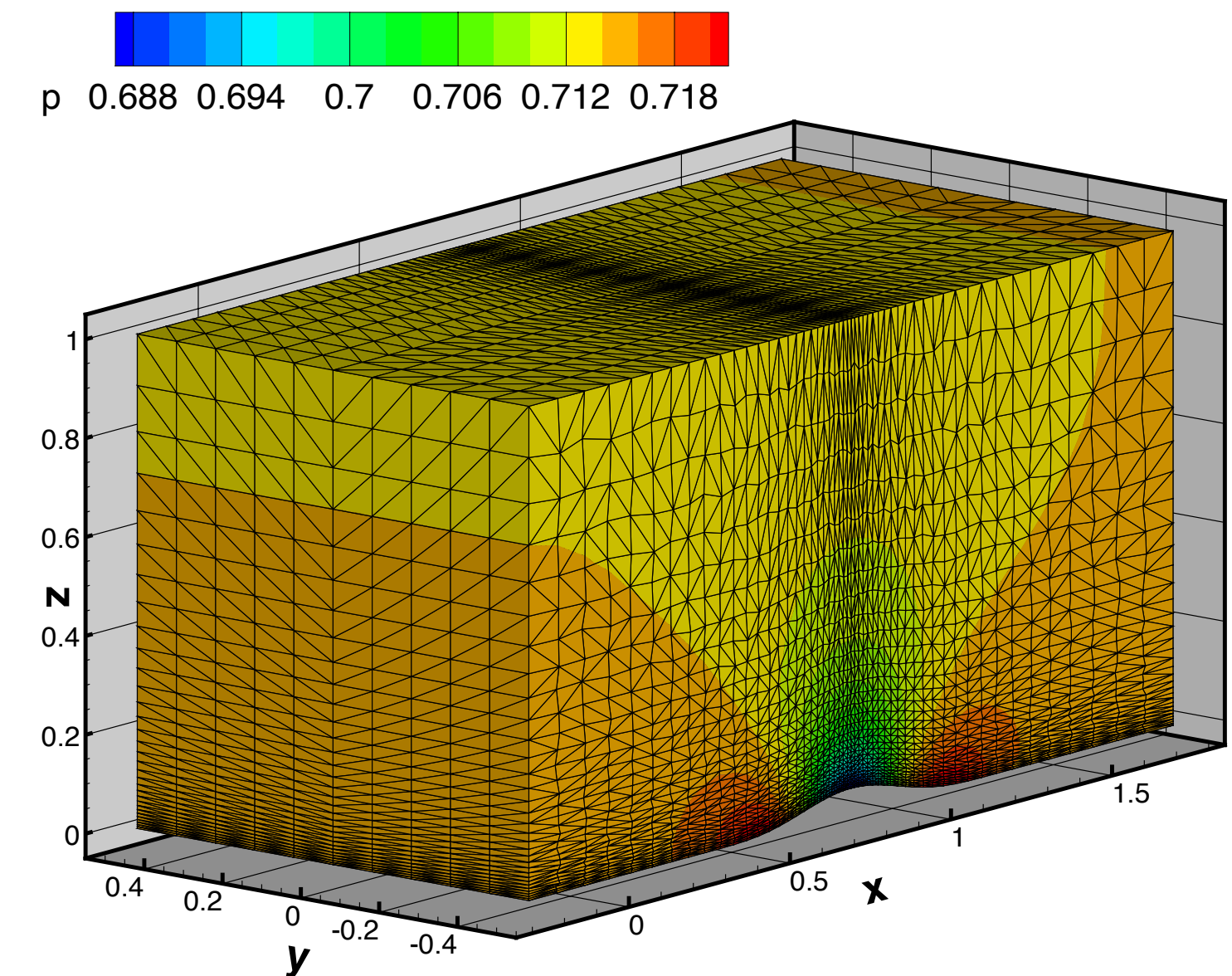
Grid 1



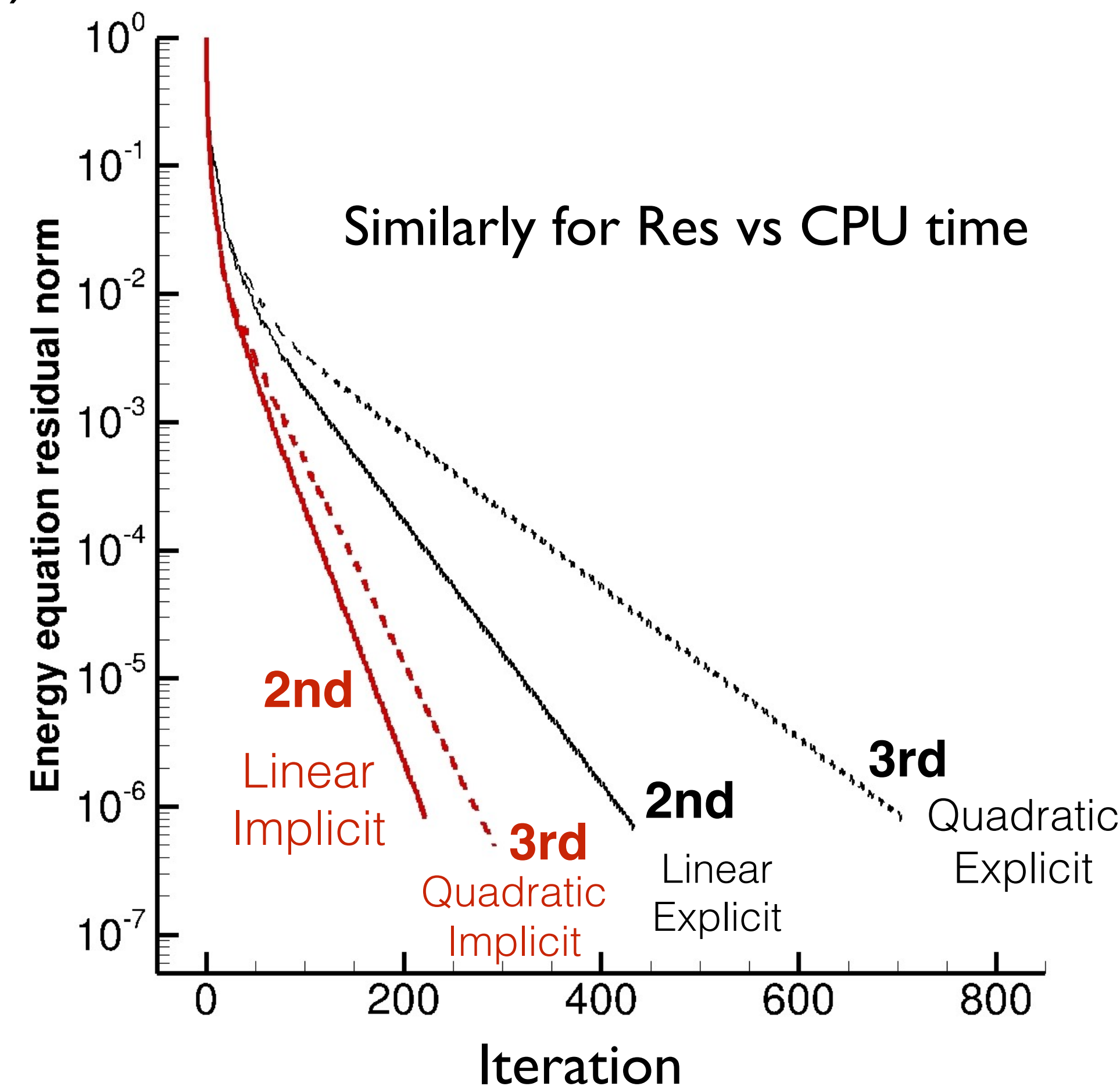
Grid 2



Grid 3

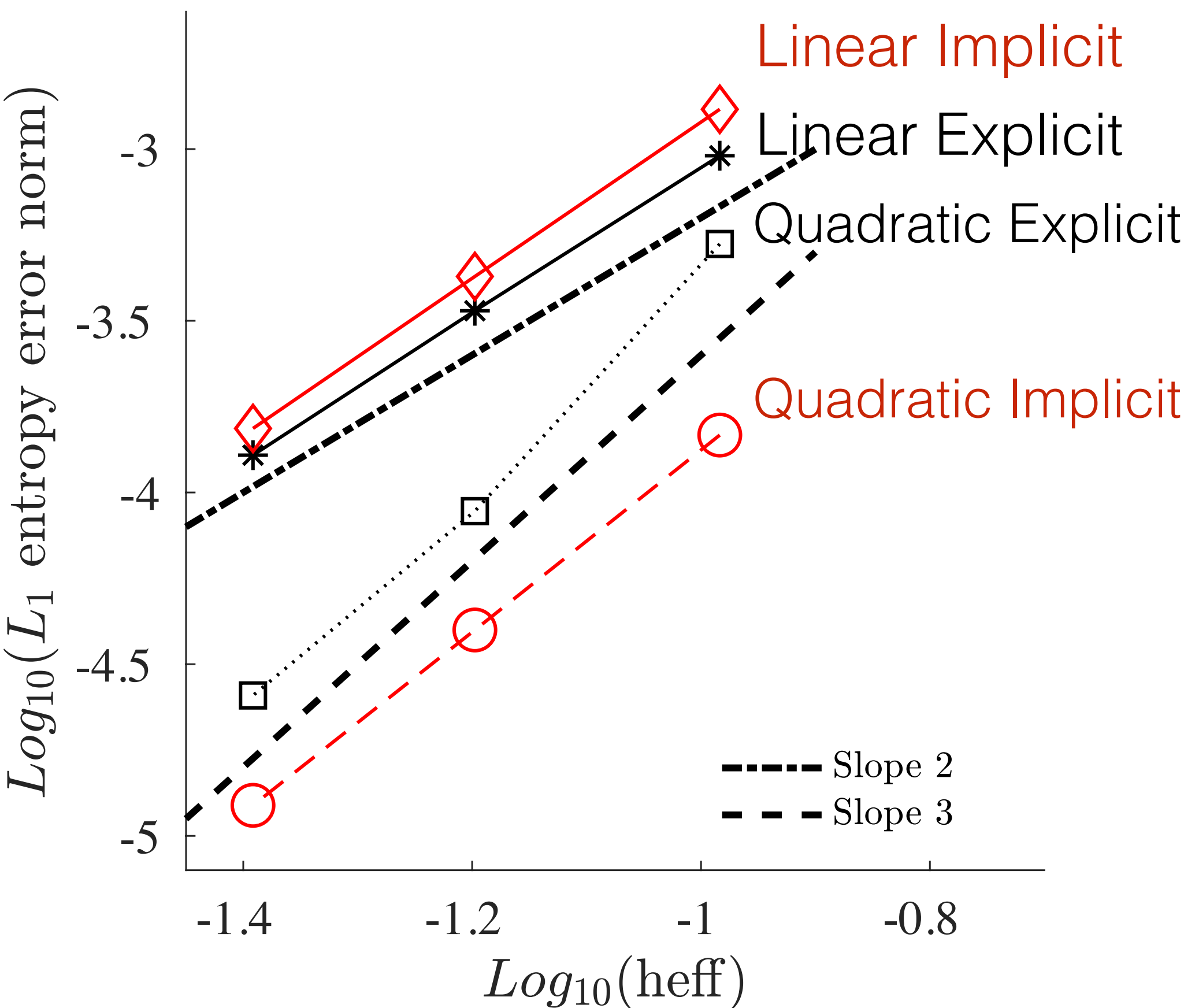


(1) Residual norm vs iteration



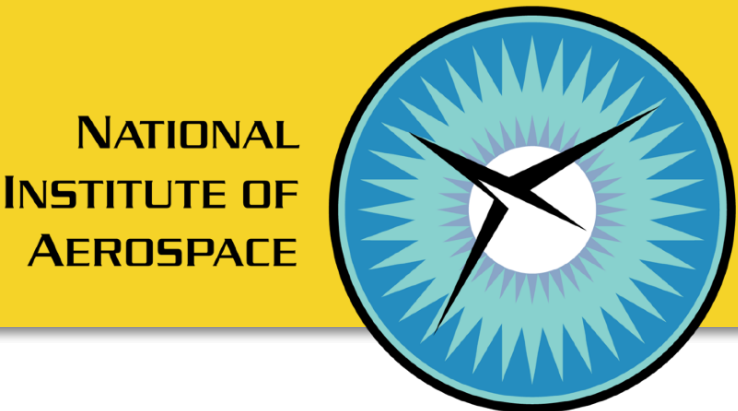
Fewer iterations and faster time-to-solution
IEBG methods are not expensive.

(2) Entropy error convergence

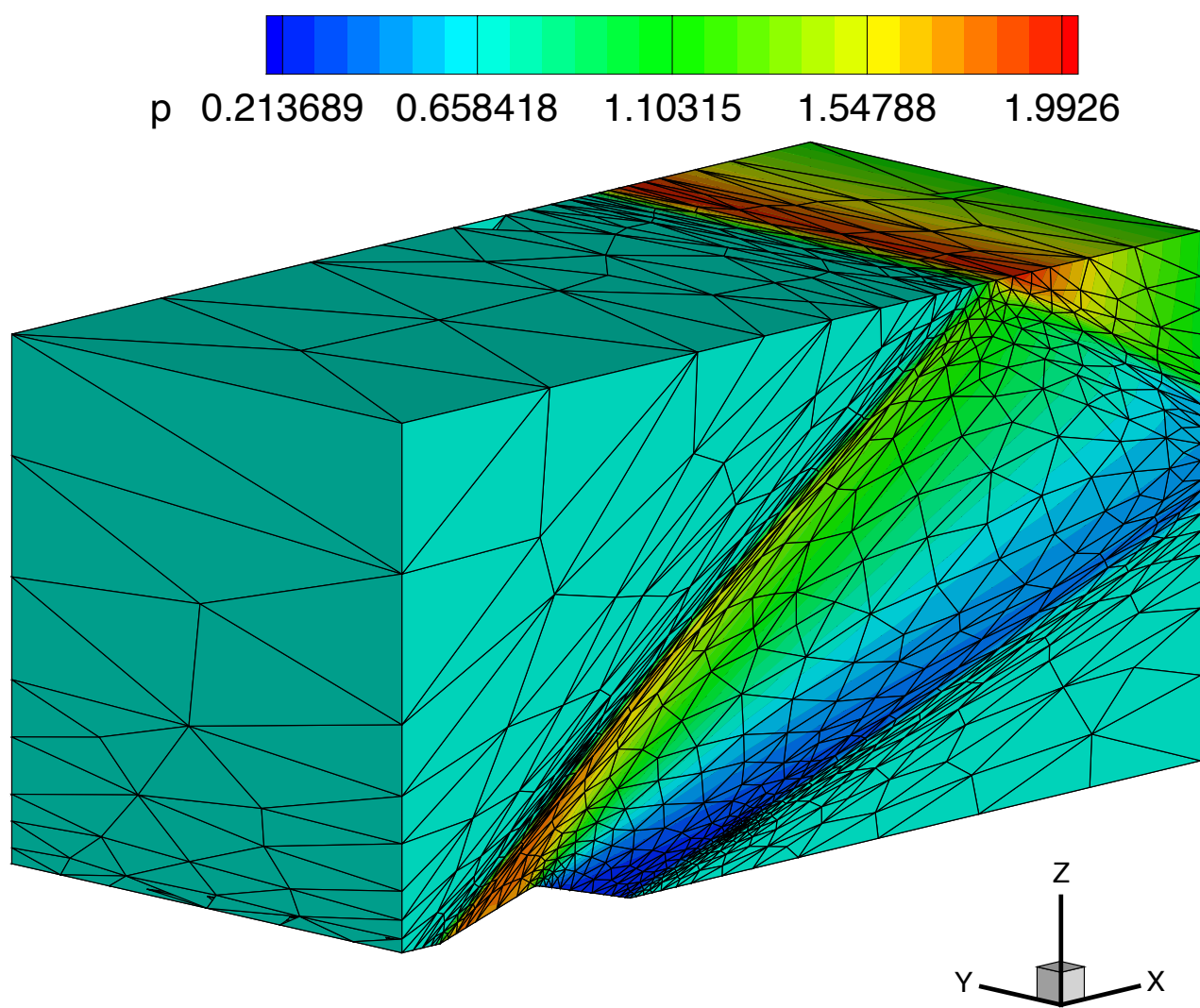


2nd- and 3rd-order convergence
Lower 3rd-order error by QIEBG.

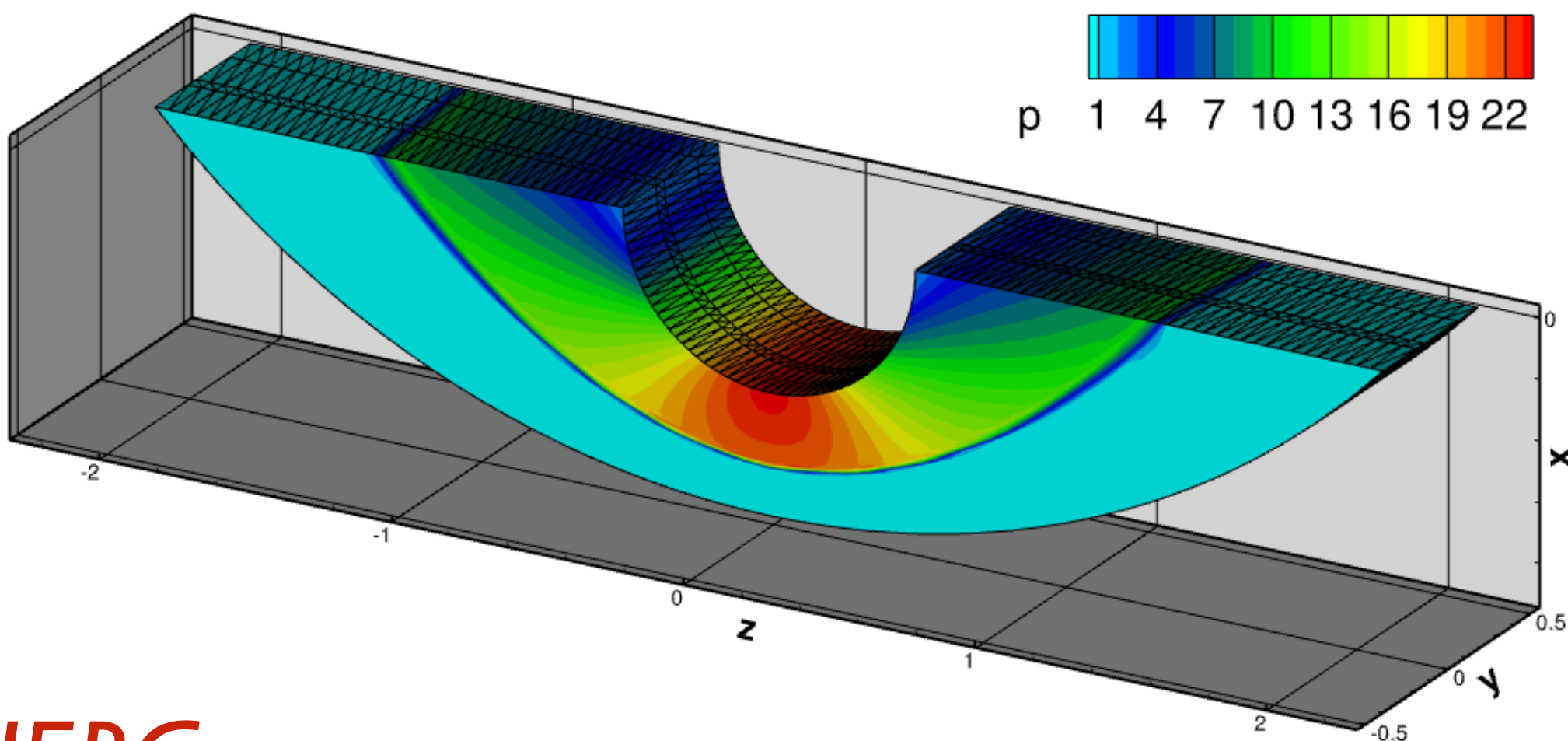
Supersonic/Hypersonic Flow: $\kappa = 0$ and $c = 5$



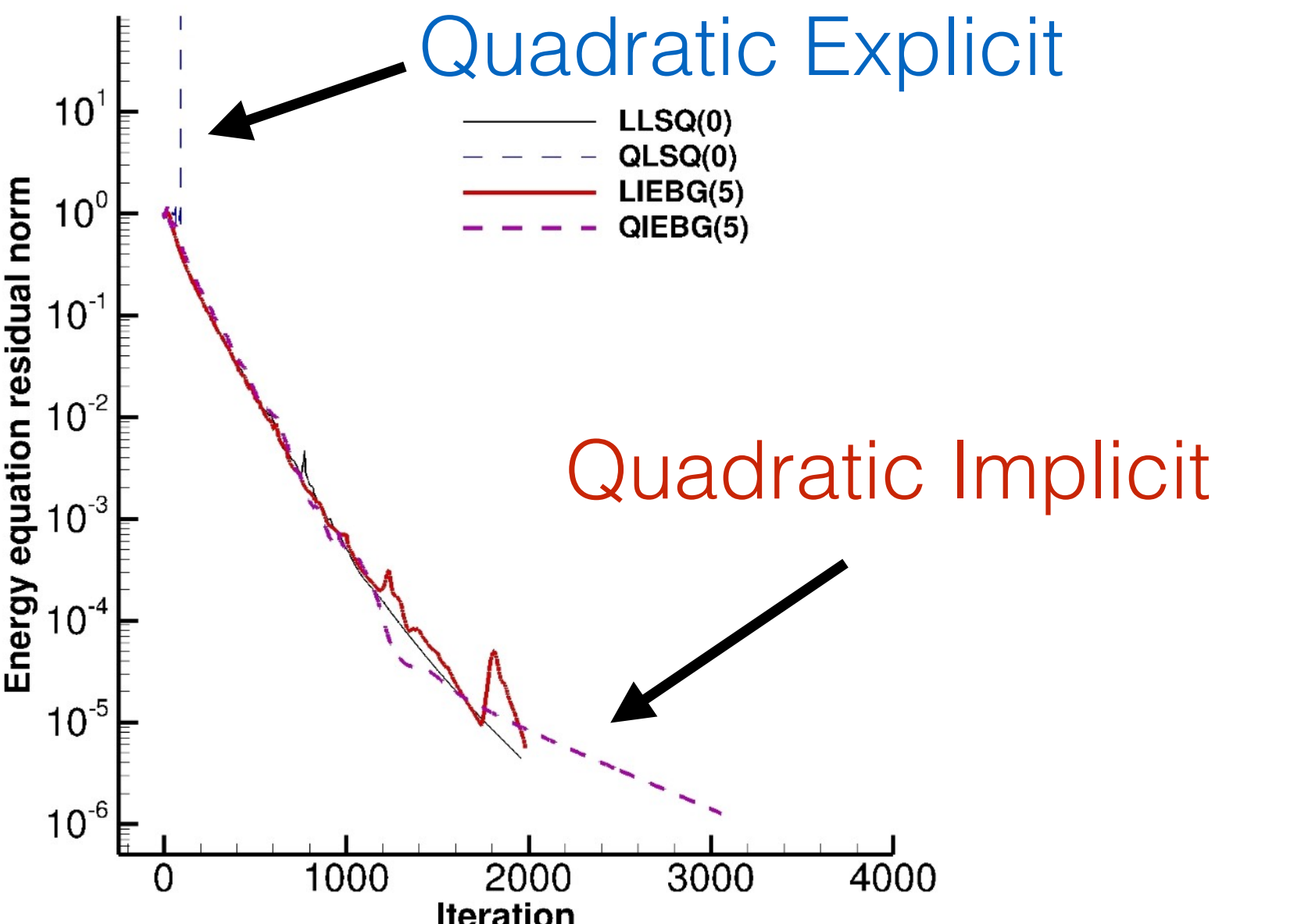
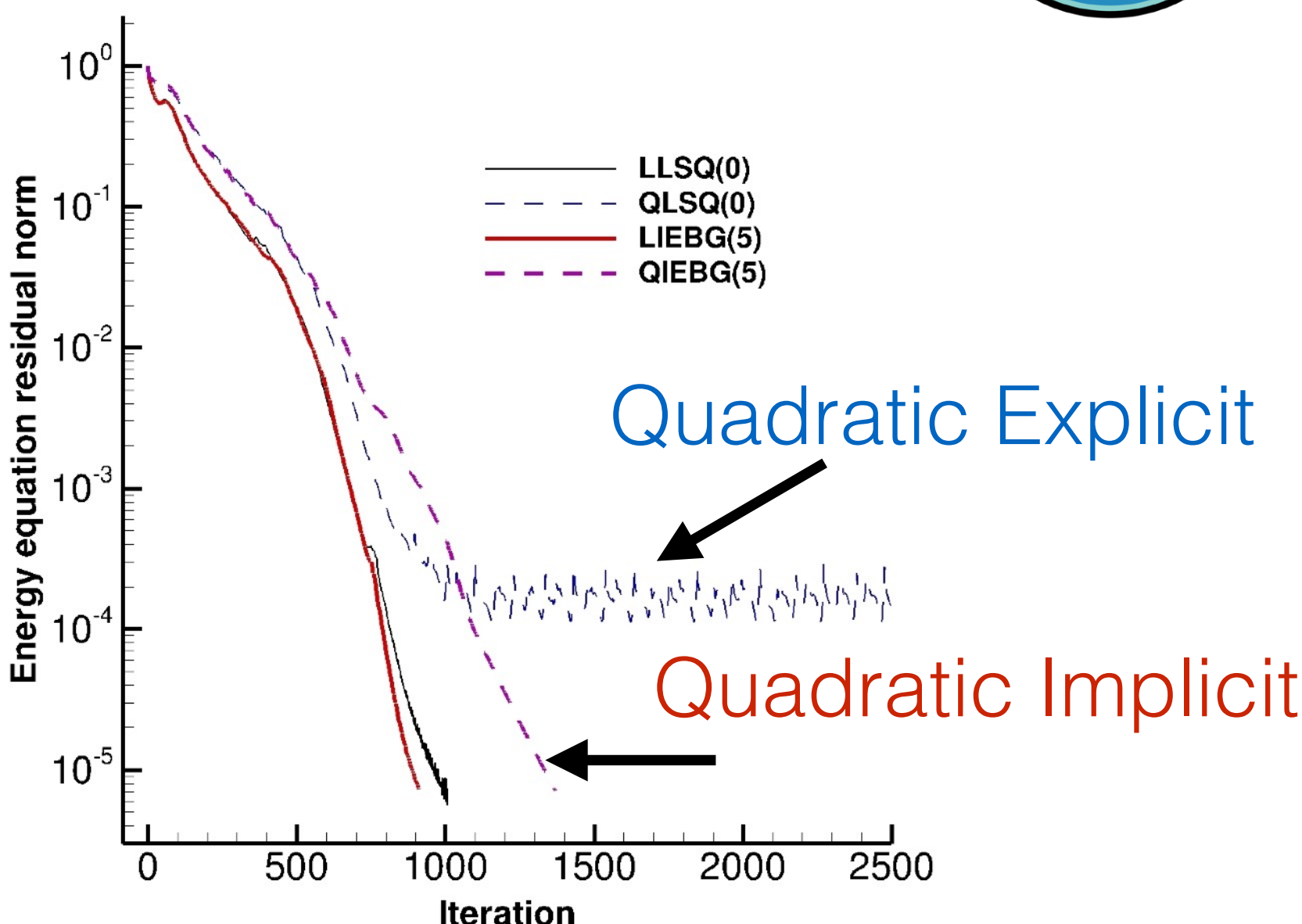
Mach 2.5 flow over a triangular bump.



Mach 5.0 flow over a circular cylinder.



More robust with QIEBG.



Conclusions

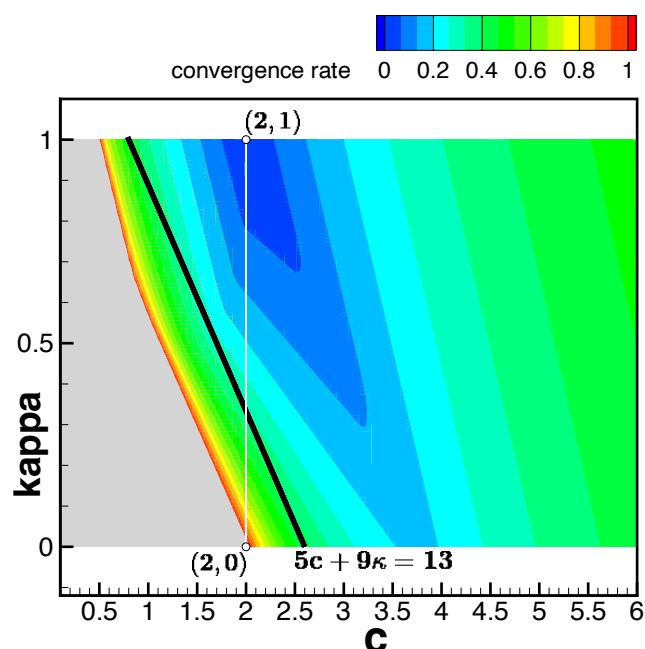
Lesson 1: *Forget that you think you know.*

2nd-order accurate gradient for any c
 $\text{Res}(c) = \nabla u - \mathbf{g} + O(h^2)$
Robust with $c = 5$.

Lesson 2: *See it for yourself.*

Simplified:
$$\mathbf{M}_{jj} = \left(\frac{c + \kappa - 1}{2} \right) \mathbf{I}$$

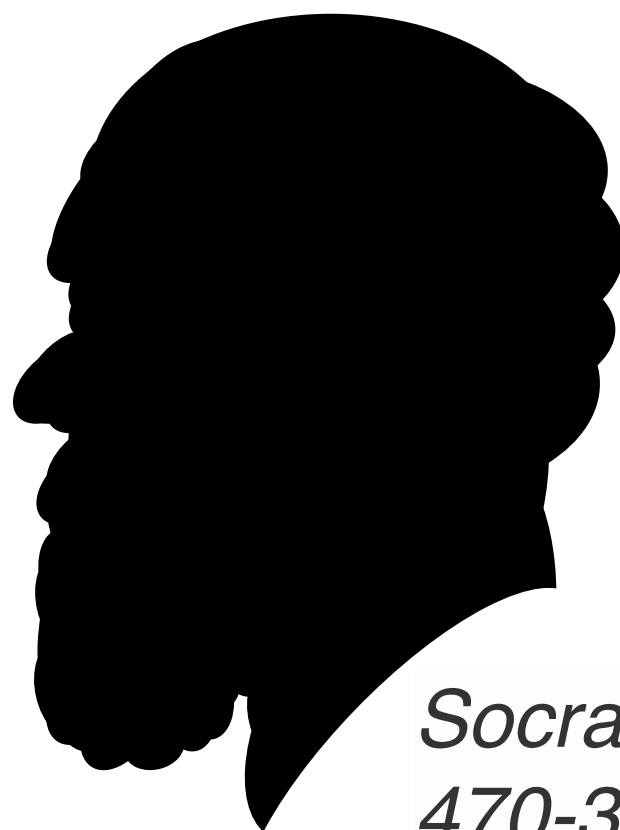
Lesson 3: *Understand limits.*



4th-order with
 $c = (13 - 9\kappa)/5$
Stable with
 $c > 2$ (QIEBG $\kappa=0$)

Results: IEBG demonstrated for subsonic/supersonic/hypersonic flows.
Lower 3rd-order errors and more robust iterative convergence than with quadratic LSQ.

Future work: Applications to space-time viscous problems and cell-centered nodal-gradient finite-volume methods.



Socrates
470-399 B.C.

“I neither know nor think that I know” (in Plato, Apology 21d).

I had to realize and admit I didn't know, or this work would have been far less complete.
3rd-order EB scheme is now made more robust.

*“The really important thing is not to live,
but to live well.”*



Socrates,
470-399 B.C.

*The really important thing is not to develop CFD algorithms,
but to develop them well.*

*Always seek a better understanding, a further simplification,
and a better guide on how to choose parameters.*