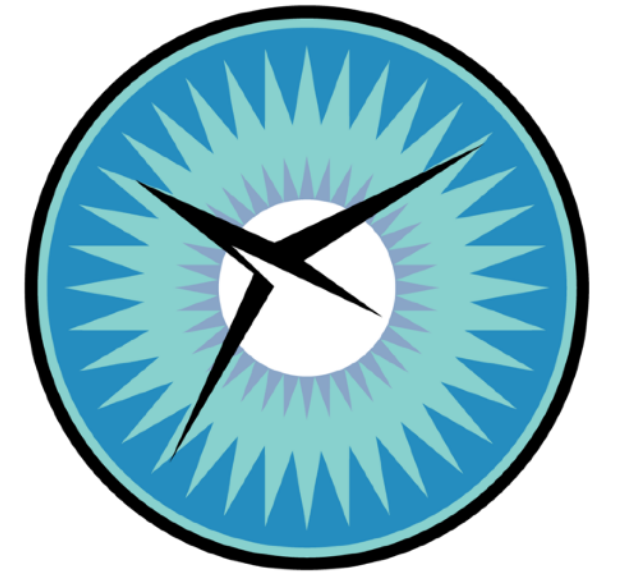


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Towards Robust and Accurate Implicit Gradient Methods for Second- and Third-Order Nodal-Gradient Cell-Centered Finite-Volume Discretizations on Tetrahedral Grids

Hiroaki Nishikawa, *National Institute of Aerospace*

*11:10am, in Academic Ballroom 412, AIAA Aviation,
Tuesday, July 30, 2024*

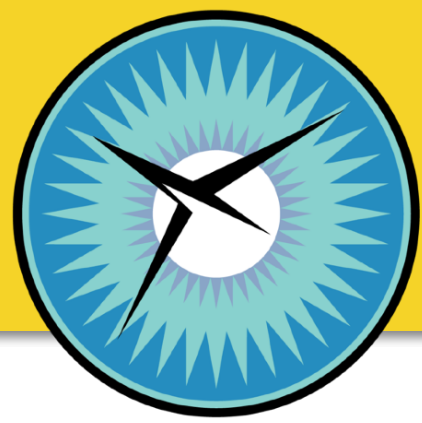


This work was supported partly by the Hypersonic Technology Project, through the Hypersonic Airbreathing Propulsion Branch of the NASA Langley Research Center, under Prime Contract No. 80LARC23DA003

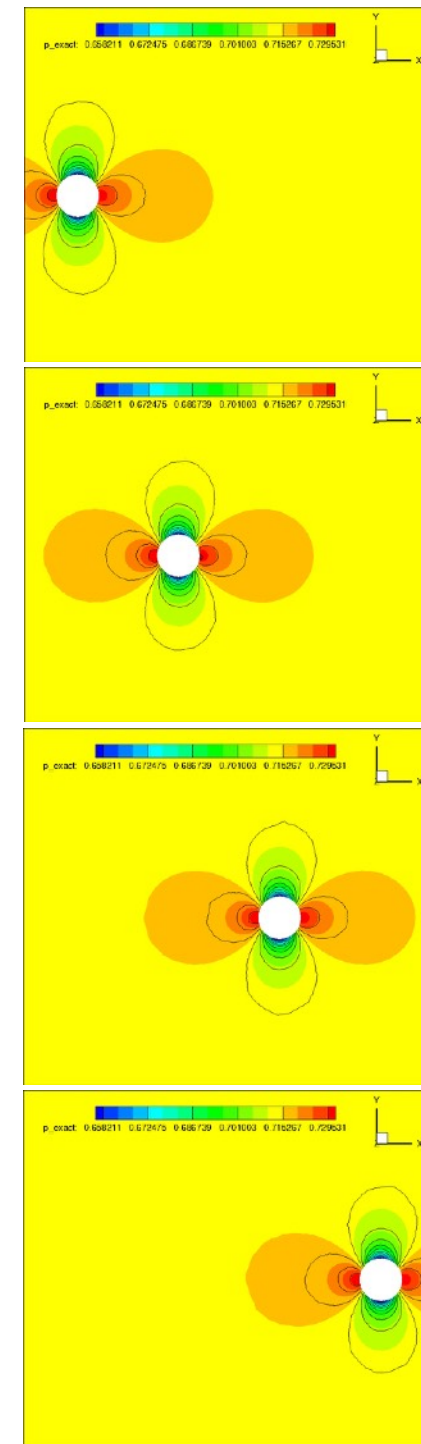
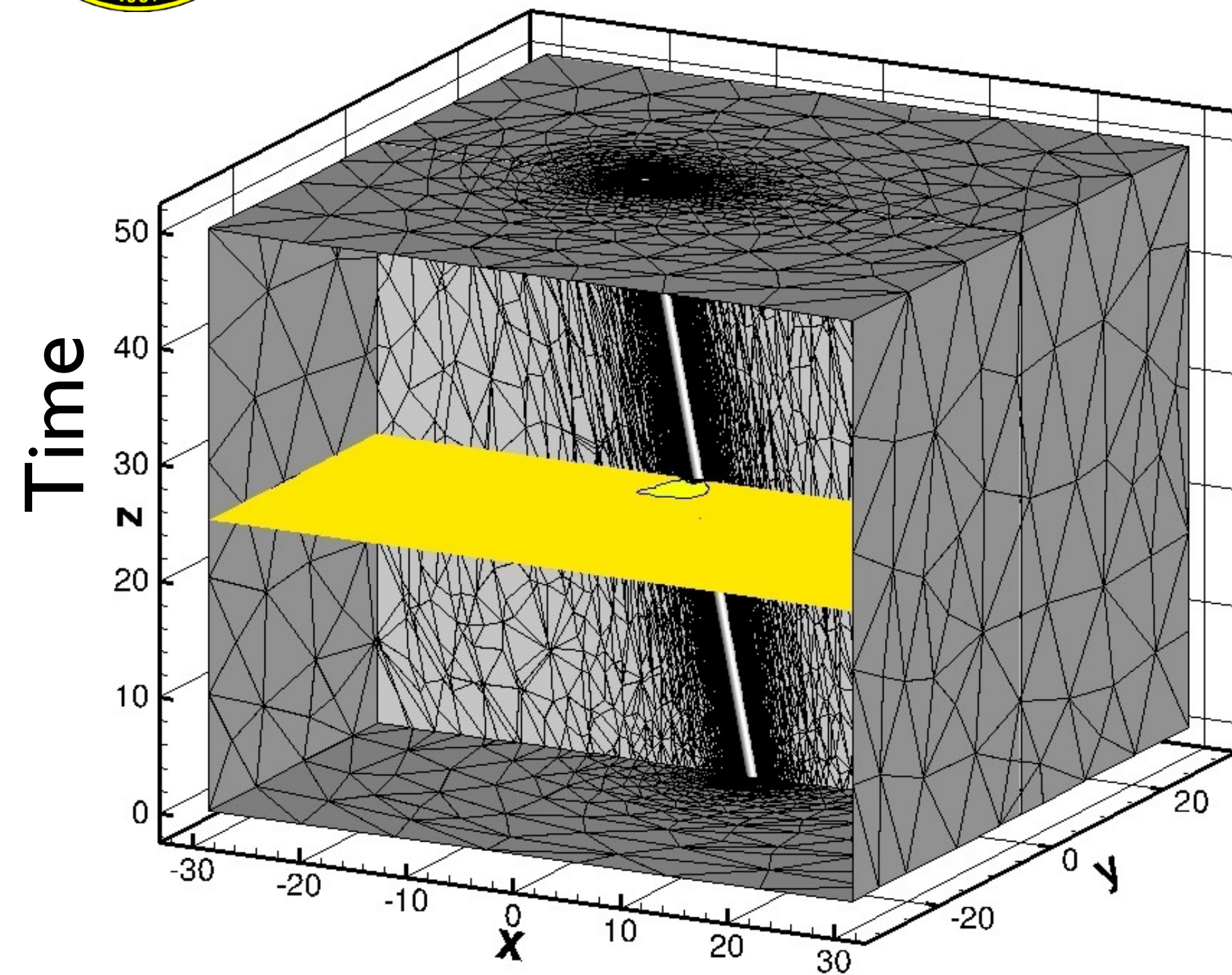
Economical 3rd-Order Methods

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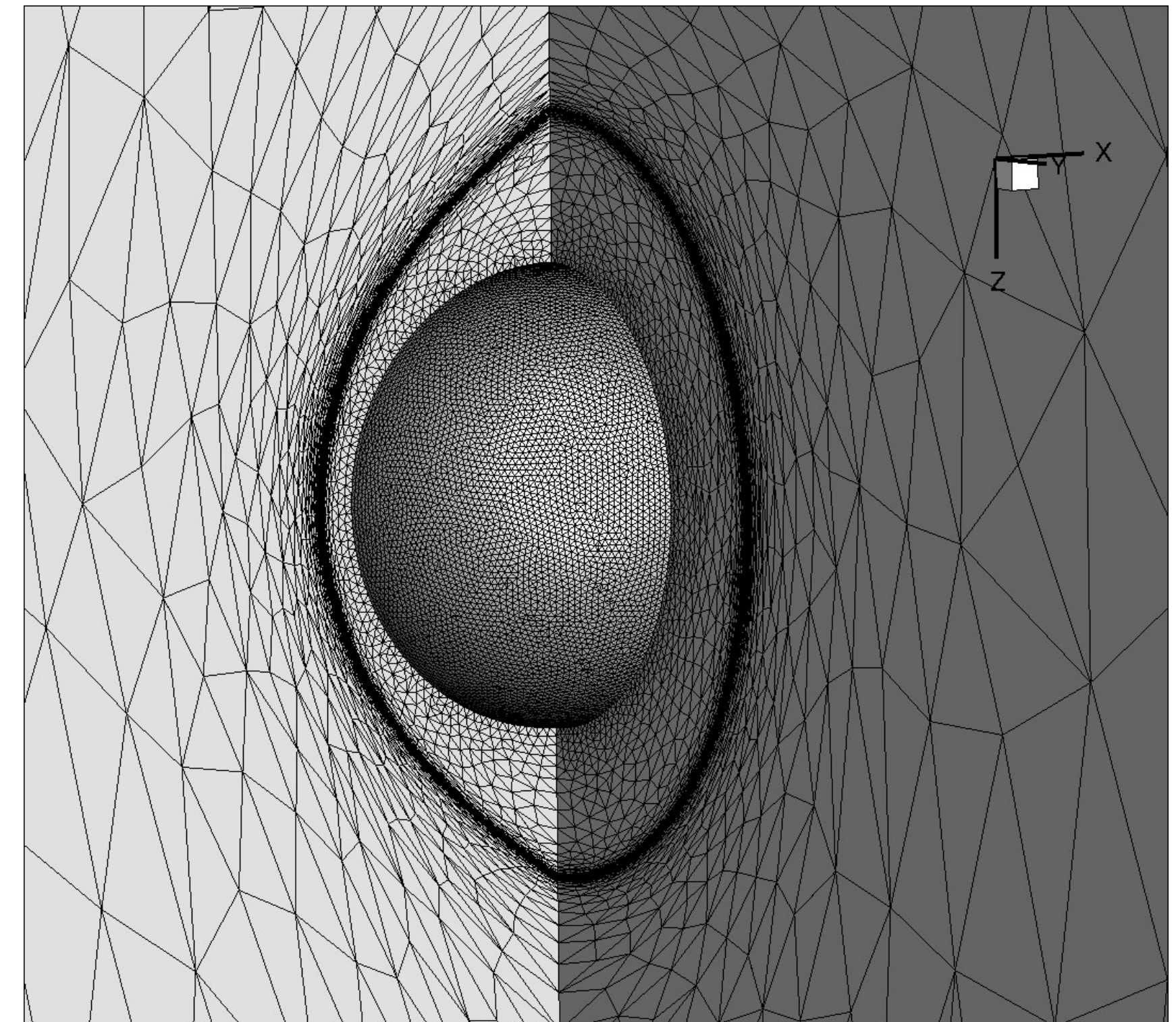
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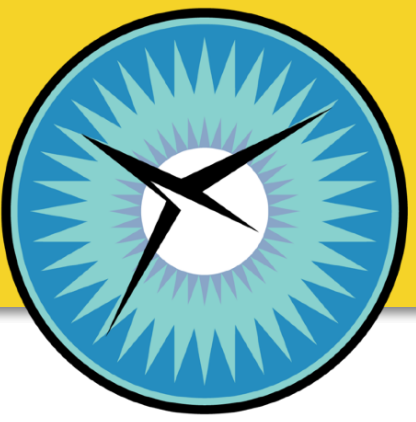
Exploring Space-Time Hyperbolic NS



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.



This paper focuses on a 3rd-order cell-centered method:

$$V_j \frac{\partial \mathbf{u}_j}{\partial t} + \sum_{k \in \{k_j\}} \Phi_{jk}(\mathbf{w}_L, \mathbf{w}_R) |\mathbf{n}_T| - \mathbf{s}_j V_j = \delta \mathbf{f}_j + \delta \mathbf{s}'_j$$

where $\mathbf{u}_j$ is a vector of point-valued solutions at cell center, and

Quadratic
interpolation:

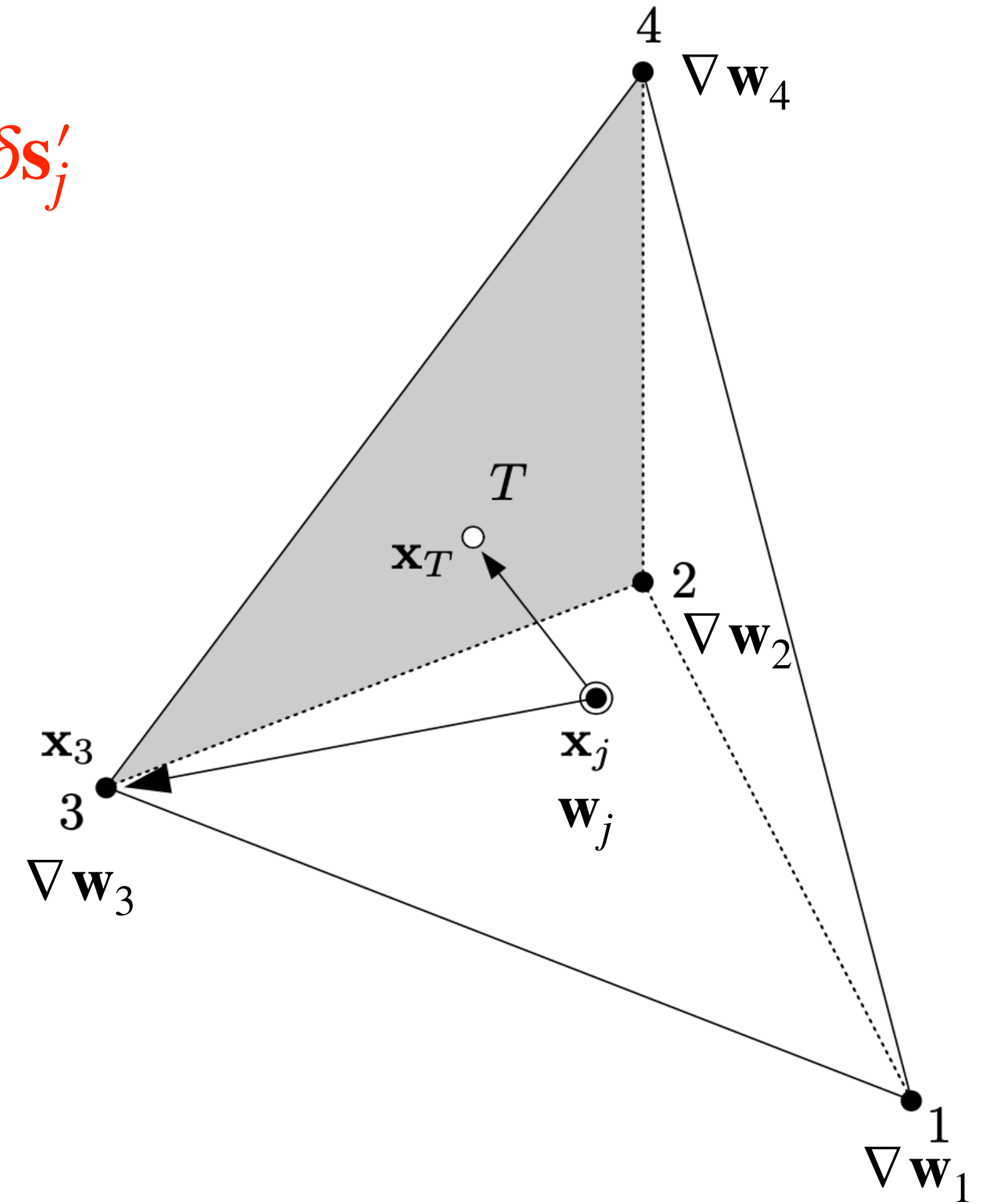
$$\mathbf{w}_L = \mathbf{w}_j + \frac{1}{2} (\overline{\nabla \mathbf{w}_T} + \overline{\nabla \mathbf{w}_j}) (\mathbf{x}_T - \mathbf{x}_j)$$

Flux correction:

$$\delta \mathbf{f}_j = \frac{1}{24} \sum_{k \in \{k_j\}} \sum_{i=1}^3 \left(\frac{\partial \mathbf{f}}{\partial \mathbf{w}} \right) \nabla \mathbf{w}_i (\mathbf{x}_i - \mathbf{x}_T) |\mathbf{n}_T|$$

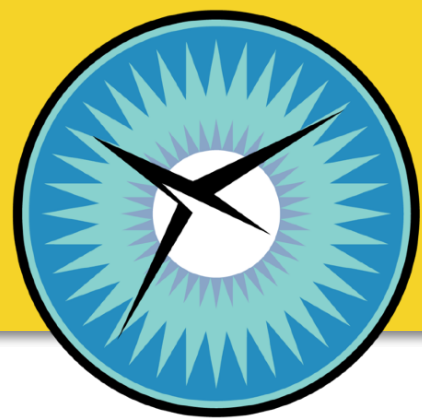
Source correction:

$$\delta \mathbf{s}'_j = \frac{1}{40} \sum_{i=1}^4 \nabla \mathbf{s}'_i (\mathbf{x}_i - \mathbf{x}_j) V_j \quad \mathbf{s}' = \mathbf{s}_j - \frac{\partial \mathbf{u}}{\partial t}$$



We only need gradients at nodes for 3rd-order accuracy: no second derivatives required.

$$\mathbf{g} = \left[\partial_x u, \partial_y u, \partial_z u \right]$$



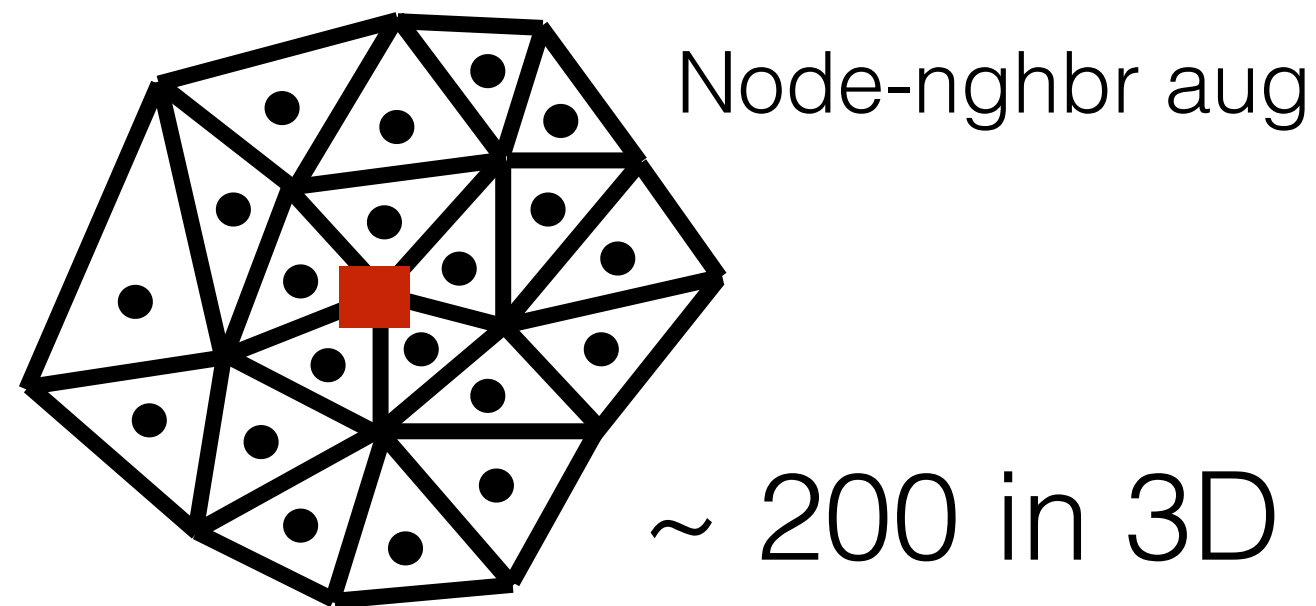
Explicit:

Linear LSQ

$$\mathbf{g}_j = \sum_{k \in \text{nghbrs}} C_{jk} u_k$$

Quadratic LSQ

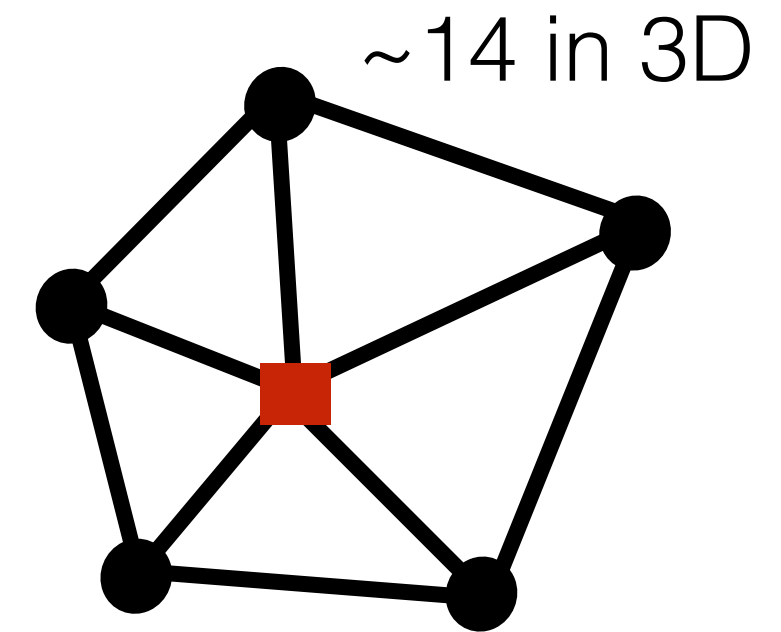
$$\mathbf{g}_j = \sum_{k \in \text{nghbrs}} C_{jk} u_k$$



Large stencil: a solver gets more robust but less accurate.
[Haider&Croisille&Courbet NM2009](#)

Implicit:

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



Galerkin: [Löhner\(1994\)](#)

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

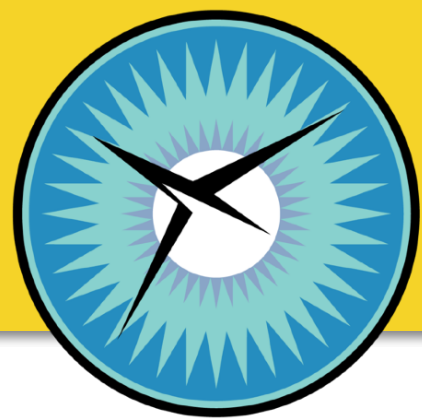
Implicit GG: [Nishikawa, JCP2019](#)

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

Implicit Gradients

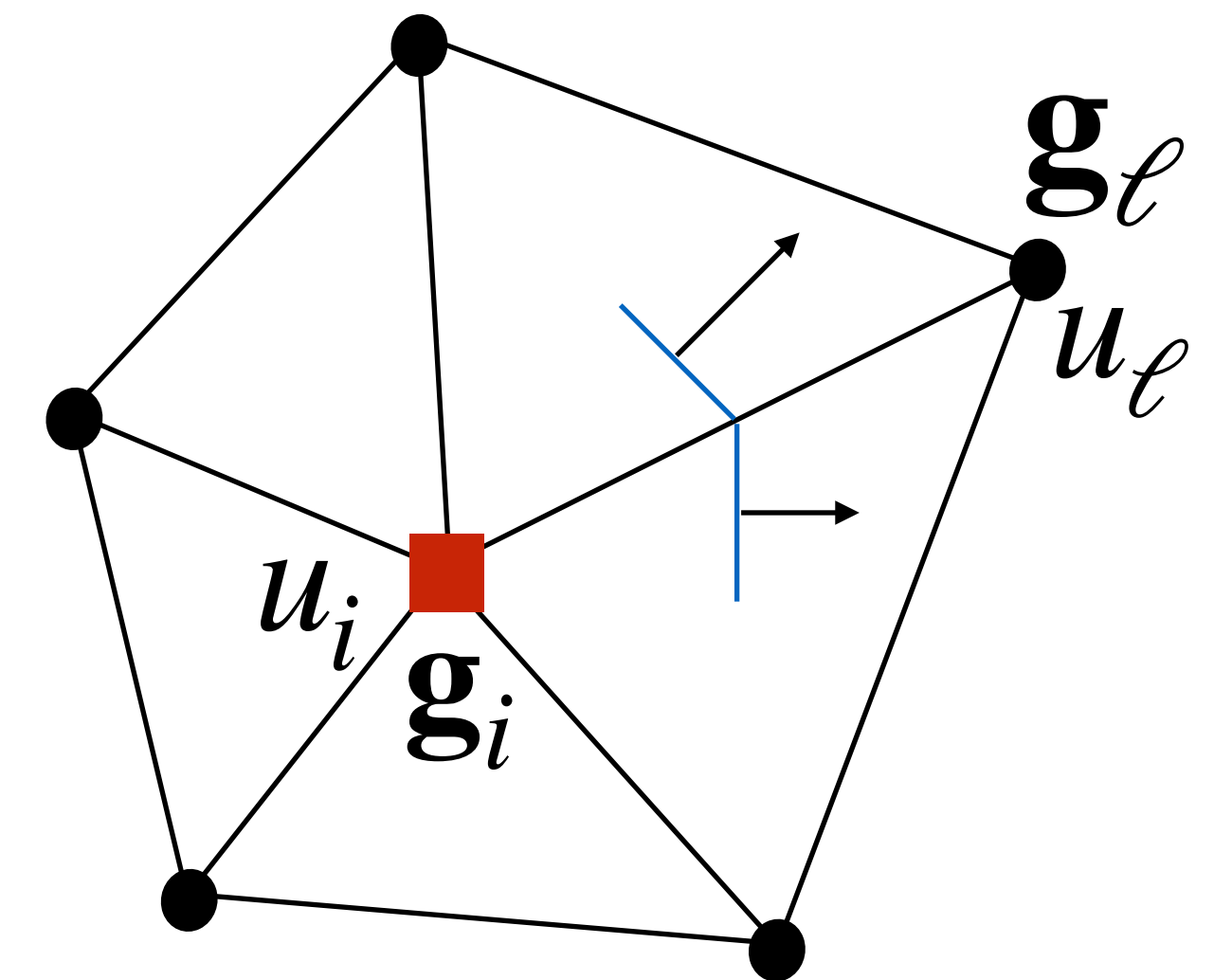


IEBG2 system: Exact for quadratic functions

$$\mathbf{M}_{ii}\mathbf{g}_i + \sum_{\ell \in \{\ell_i\}} \mathbf{M}_{i\ell}\mathbf{g}_\ell = \frac{1}{V_j} \sum_{\ell \in \{\ell_i\}} \frac{u_i + u_\ell}{2} \mathbf{n}_{i\ell} \quad \mathbf{g}_i = \begin{bmatrix} \partial_x u_i \\ \partial_y u_i \\ \partial_z u_i \end{bmatrix}$$

We can solve the system by a relaxation scheme:

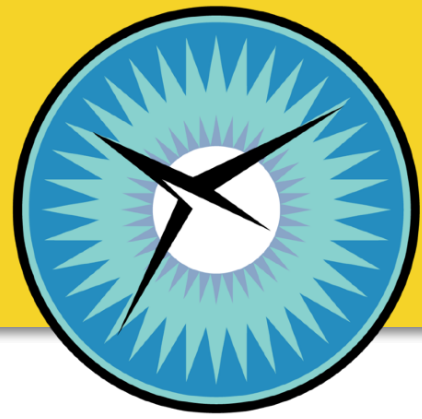
$$\mathbf{g}_i^{m+1} = \mathbf{g}_i^m - \mathbf{M}_{ii}^{-1} \left(\sum_{\ell \in \{\ell_i\}} \mathbf{M}_{i\ell}\mathbf{g}_\ell^m + \frac{1}{V_j} \sum_{\ell \in \{\ell_i\}} \frac{u_i + u_\ell}{2} \mathbf{n}_{i\ell} \right)$$



Successfully demonstrated for their 3rd-order EB discretization method.

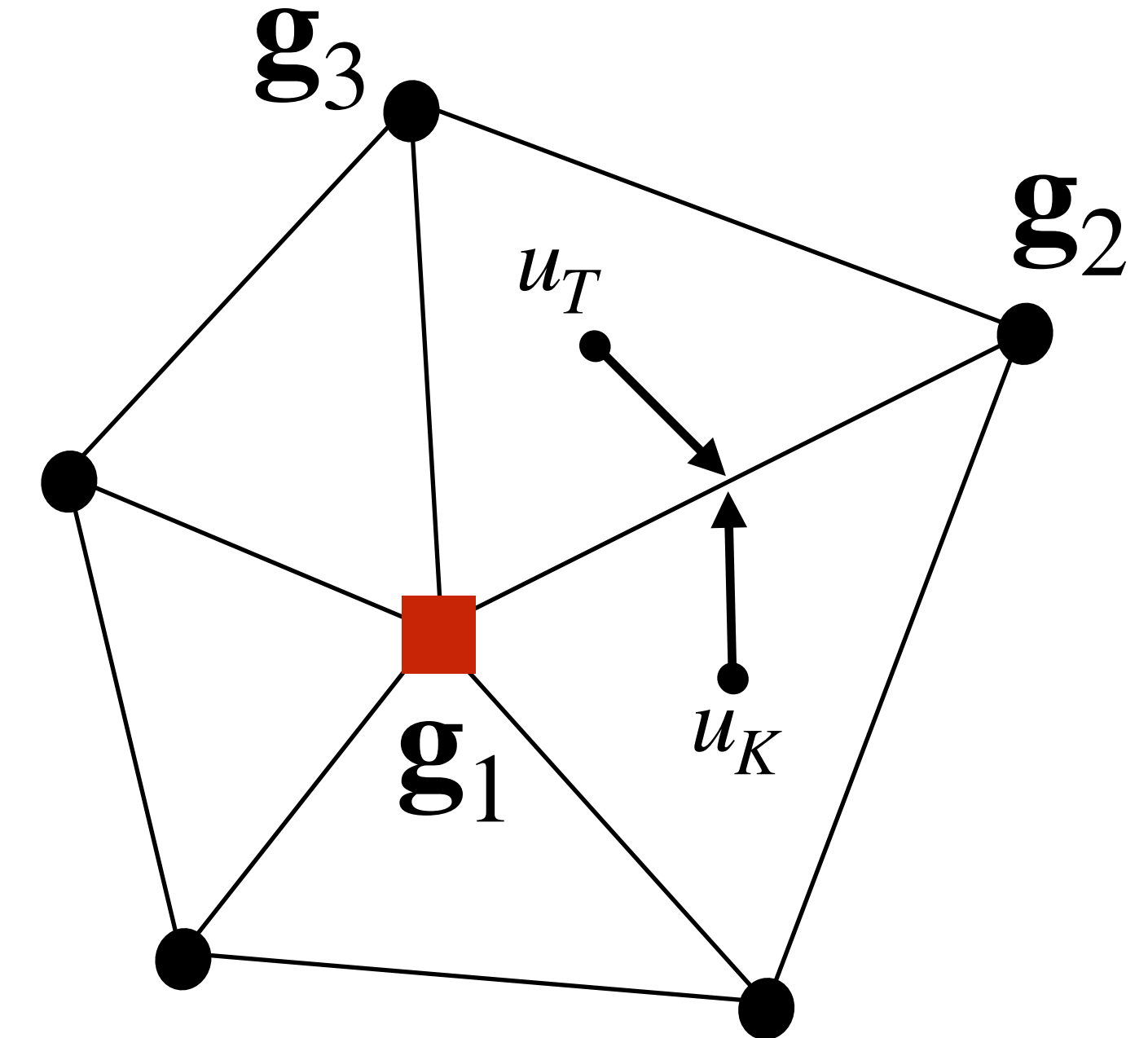
IEBG2 is very efficient and perfect for the 3rd-order EB because it does not require second derivatives.

Can we develop something similar to our 3rd-order CCFV method?



Yes, we can. First, define the cell discretization:

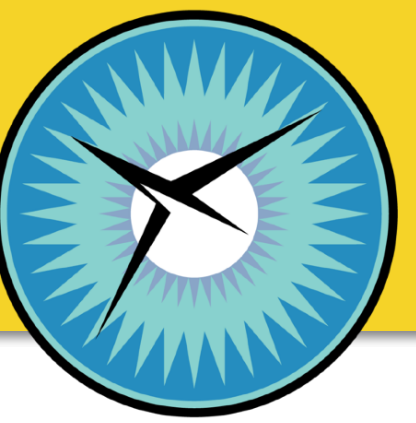
$$\mathbf{g} = \nabla u \longrightarrow \int_{V_T} \mathbf{g} dV = \oint_{\partial V_T} u \hat{\mathbf{n}} ds \xrightarrow{\text{3rd-order NG-CCFV, No 2nd derivatives}} \mathbf{Res}_T = \frac{\mathbf{g}_1 + \mathbf{g}_2 + \mathbf{g}_3}{3} - \sum_{K \in \{K_T\}} [\phi_{TK} + \delta\phi_{TK}] |\mathbf{n}_{TK}|$$



Then, minimizing the cell residuals in the LSQ sense:

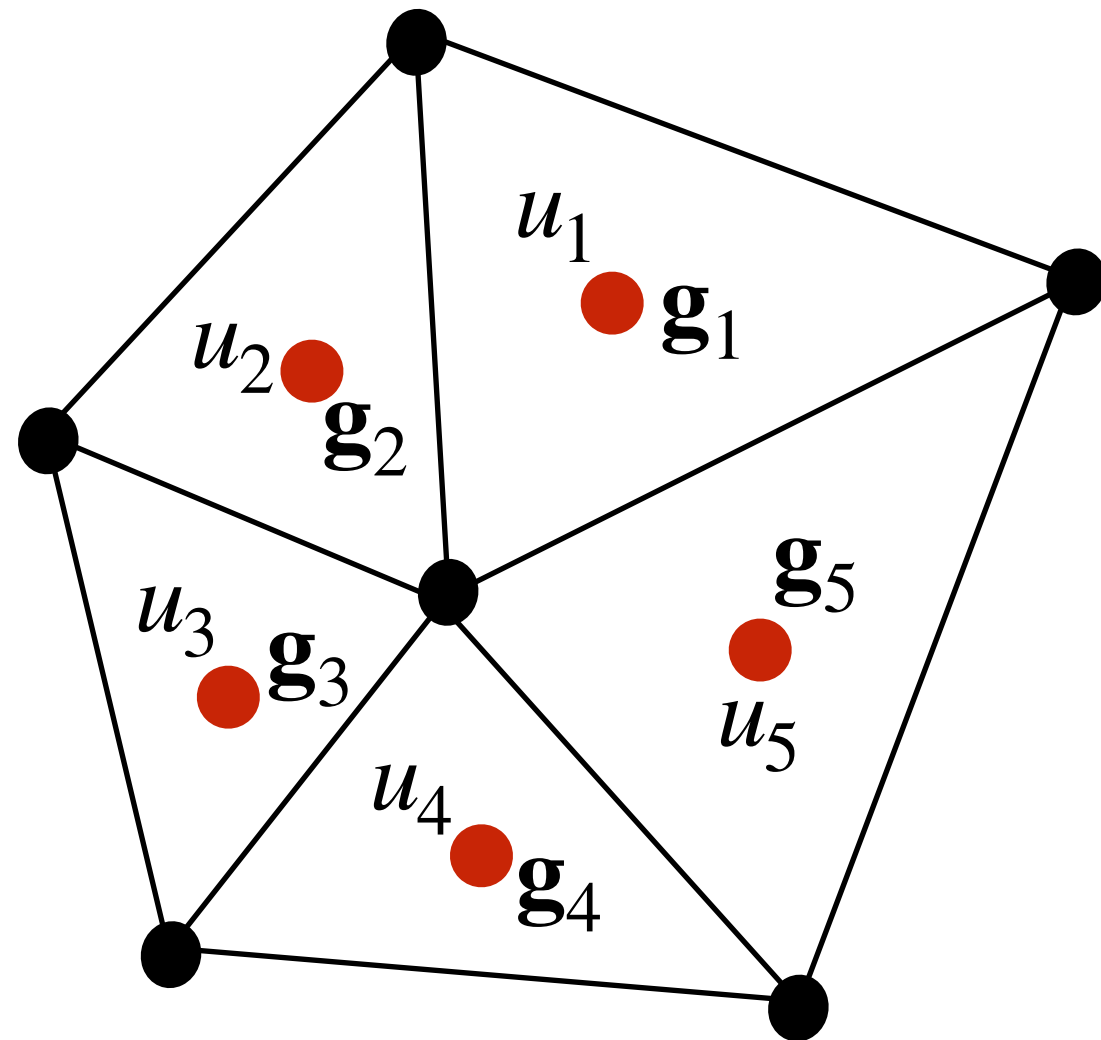
$$\mathcal{F} = \frac{1}{2} \sum_{\text{all cells}} \mathbf{Res}_T^2 \longrightarrow \frac{\partial \mathcal{F}}{\partial \mathbf{g}_i} \equiv \mathbf{Res}_i = \sum_{T \in \{T_i\}} \left(\frac{\partial \mathbf{Res}_T}{\partial \mathbf{g}_i} \right)^t \mathbf{Res}_T = \mathbf{0}$$

*The resulting system is second-derivative-free as desired, but not robust...
Worked very hard to make it robust, but couldn't get it done successfully...*

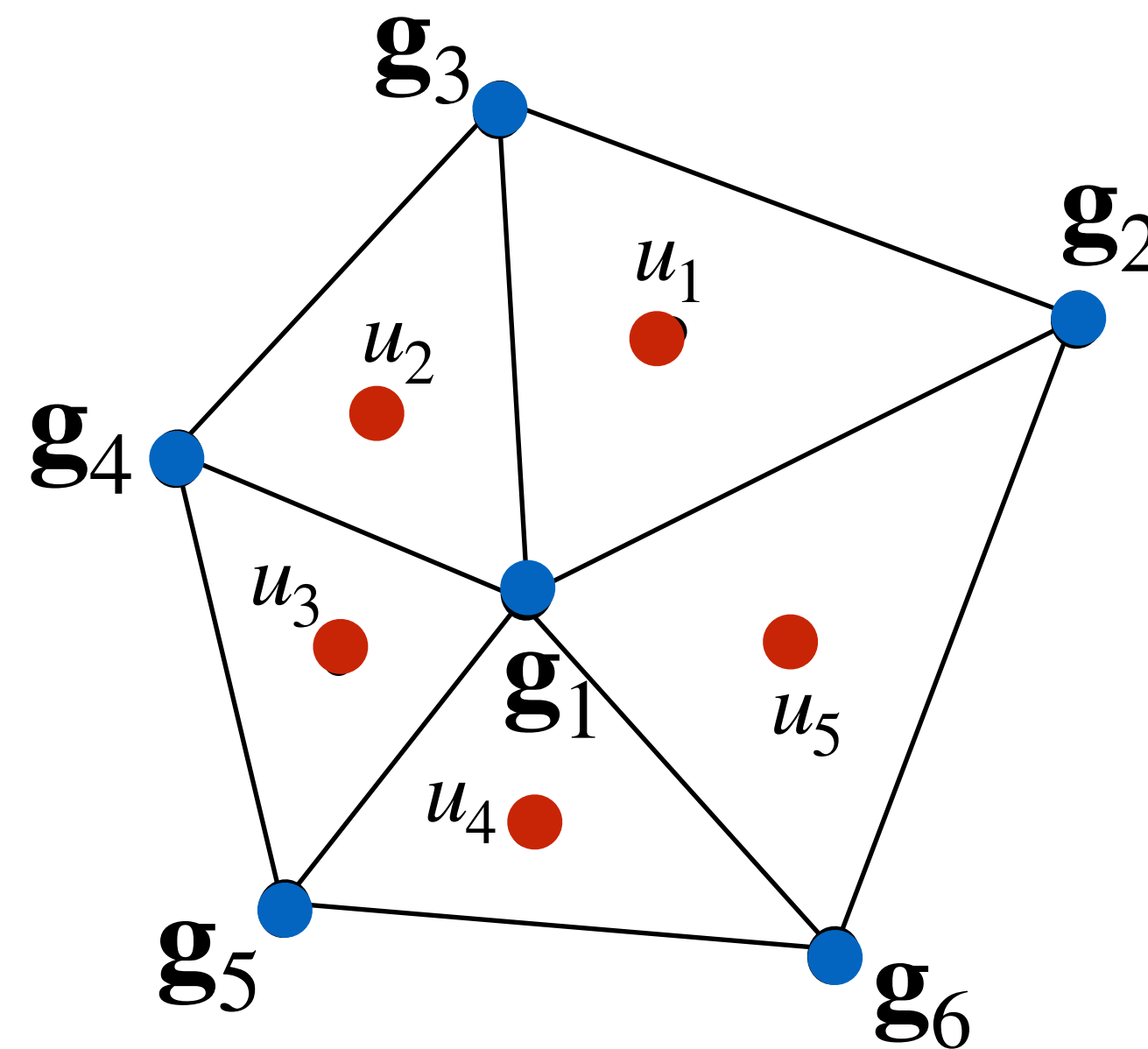


Realize that NG-CCFV is somewhere between CCFV and NCEB:

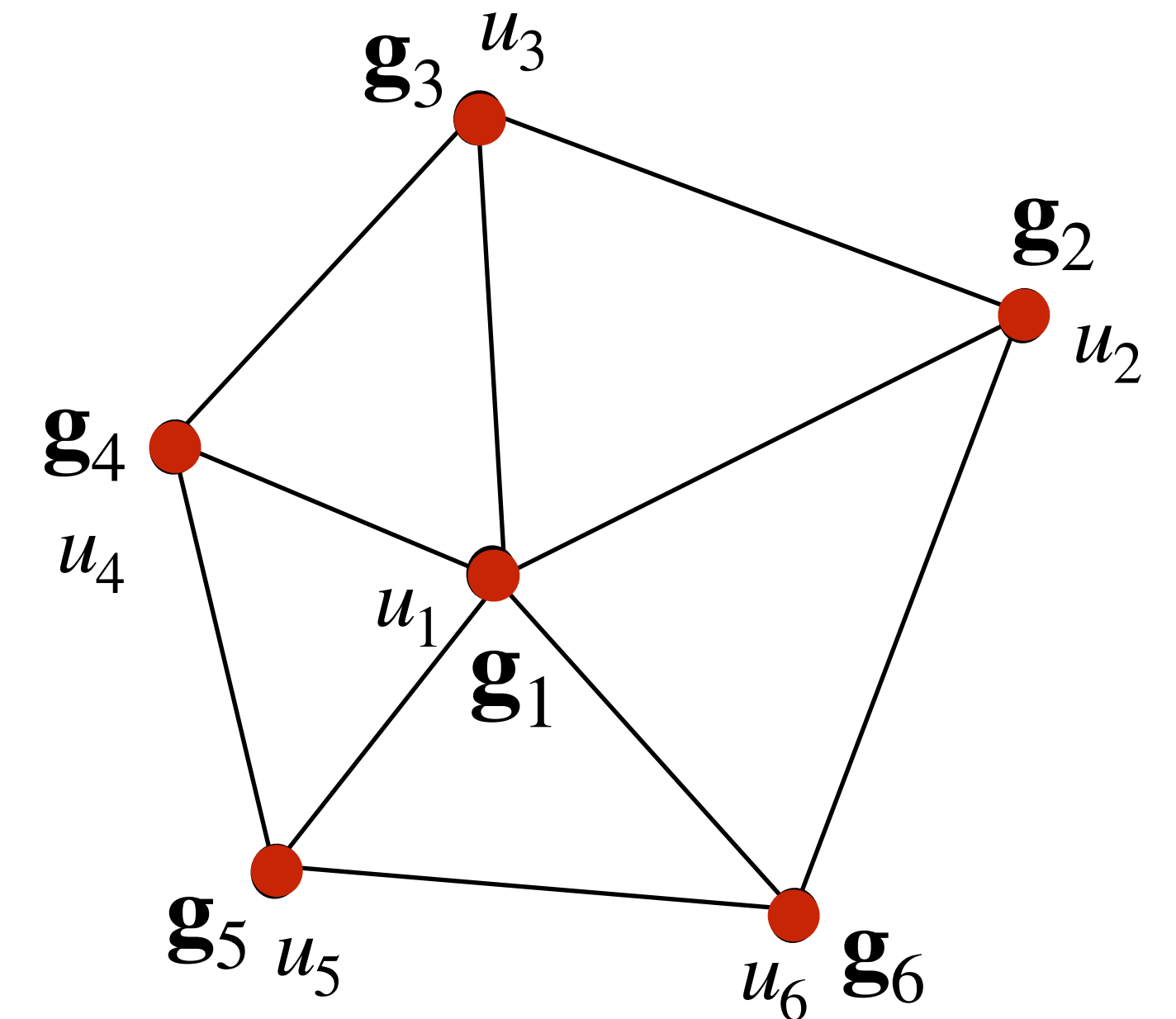
Cell-Centered



Nodal-Grad Cell-Centered

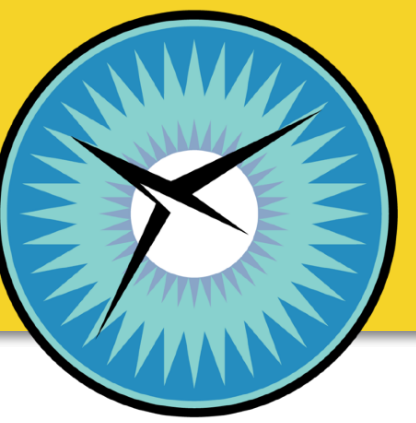


Node-Centered



$$\mathbf{M}_{ii}\mathbf{g}_i + \sum_{\ell \in \{\ell_i\}} \mathbf{M}_{i\ell}\mathbf{g}_\ell = \frac{1}{V_j} \sum_{\ell \in \{\ell_i\}} \frac{u_i + u_\ell}{2} \mathbf{n}_{i\ell}$$

IEBG2 can be directly employed if the solution values can be transferred from cells to nodes.



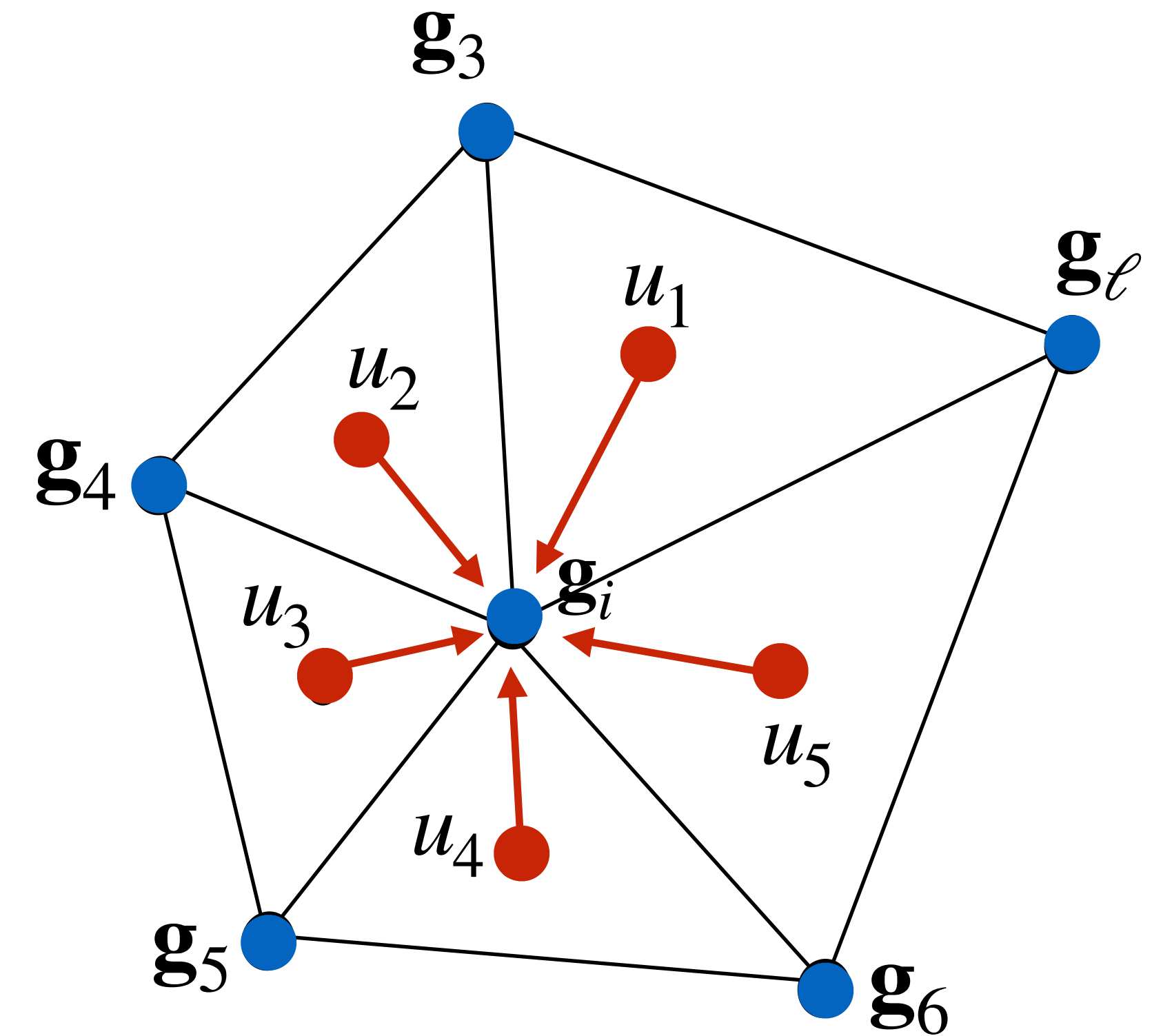
Just interpolate the solution from cells to nodes:

$$u_i = \frac{1}{N_{c_i}} \sum_{j \in \{c_i\}} \left[u_j + \left\{ C \mathbf{g}_i + (1 - C) \bar{\mathbf{g}}_j \right\} \cdot (\mathbf{x}_i - \mathbf{x}_j) \right]$$

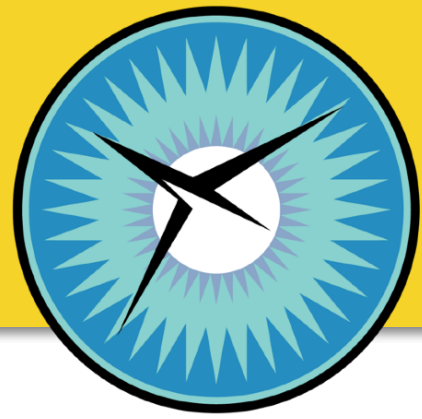
where $C=1/2$ for quadratic interpolation, otherwise linear.

Then, IEBG2 can be directly employed:

$$\mathbf{M}_{ii} \mathbf{g}_i + \sum_{\ell \in \{\ell_i\}} \mathbf{M}_{i\ell} \mathbf{g}_\ell = \frac{1}{V_j} \sum_{\ell \in \{\ell_i\}} \frac{u_i + u_\ell}{2} \mathbf{n}_{i\ell}$$



Right hand side depends on the nodal gradients —> Defect correction iteration.

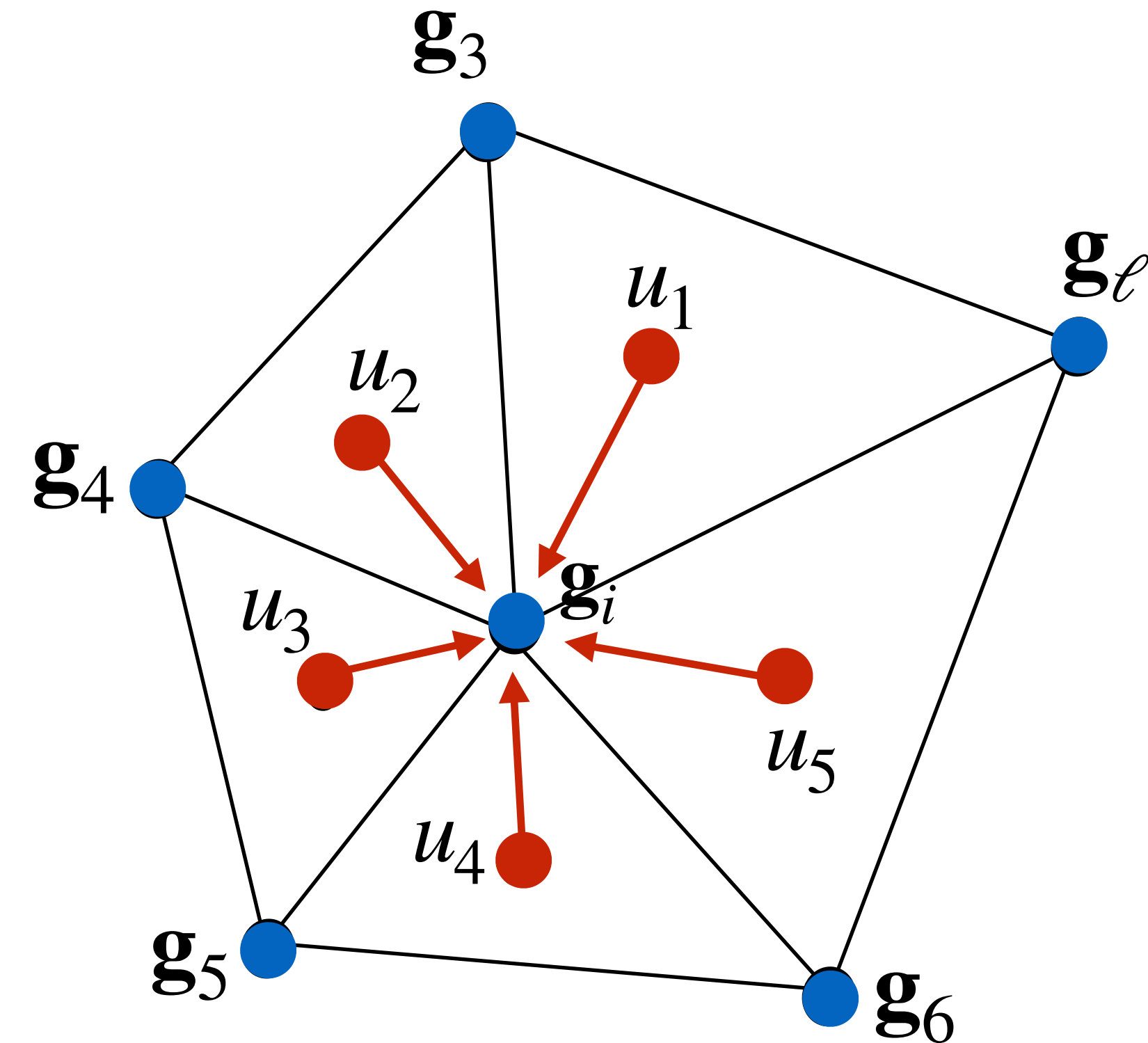


Define the residual:

$$\mathbf{r}_i = \mathbf{M}_{ii}\mathbf{g}_i + \sum_{\ell \in \{\ell_i\}} \mathbf{M}_{i\ell}\mathbf{g}_\ell - \frac{1}{V_j} \sum_{\ell \in \{\ell_i\}} \frac{u_i + u_\ell}{2} \mathbf{n}_{i\ell} \quad u_i = \frac{1}{N_{c_i}} \sum_{j \in \{c_i\}} \left[u_j + \left\{ C\mathbf{g}_i + (1 - C)\bar{\mathbf{g}}_j \right\} \cdot (\mathbf{x}_i - \mathbf{x}_j) \right]$$

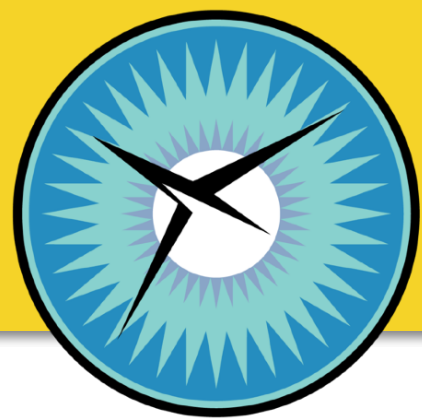
Then, we can solve it by the defect-correction iteration:

$$\begin{aligned} \mathbf{g}_i^{n+1} &= \mathbf{g}_i^n + \Delta \mathbf{g}_i \\ \Delta \mathbf{g}_i^{m+1} &= \Delta \mathbf{g}_i^m - \mathbf{M}_{ii}^{-1} \left(\sum_{\ell \in \{\ell_i\}} \mathbf{M}_{i\ell} \Delta \mathbf{g}_\ell^m + \mathbf{r}_i \right) \quad m = 0, 1, 2, 3, 4 \\ &\text{5 preconditioner relaxations (Gauss-Seidel)} \end{aligned}$$



This algorithm can be easily implemented if IEBG is already available.

Implicit Solver for Euler



Starting from the initial values:

$$\mathbf{g}_i^0 = (0,0,0), \quad i = 1,2,3,\dots,N_{nodes} \quad \mathbf{w}_j^0 = (\rho_\infty, u_\infty, v_\infty, w_\infty, p_\infty), \quad j = 1,2,3,\dots,N_{cells}$$

Then, iterate for $n = 0,1,2,3,\dots$, until convergence,

Update gradients at all nodes $i=1,2,3,\dots$, for all variables: $\mathbf{g} = (\nabla \rho, \nabla u, \nabla v, \nabla w, \nabla p)$

$$\mathbf{g}_i^{n+1} = \mathbf{g}_i^n + \Delta \mathbf{g}_i \quad \Delta \mathbf{g}_i^{m+1} = \Delta \mathbf{g}_i^m - \mathbf{M}_{ii}^{-1} \left(\sum_{\ell \in \{\ell_i\}} \mathbf{M}_{i\ell} \Delta \mathbf{g}_\ell^m + \mathbf{r}_i \right) \quad m = 0,1,2,3,4$$

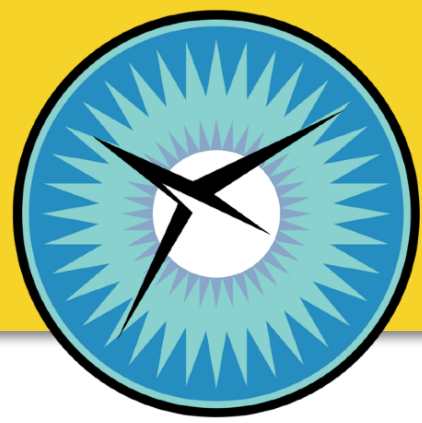
Update solutions:

$$\mathbf{U}^{n+1} = \mathbf{U}^n + \Delta \mathbf{U} \quad \left(\mathbf{D} + \frac{\partial \overline{\mathbf{Res}}}{\partial \mathbf{U}} \right) \Delta \mathbf{U} = -\mathbf{Res}(\mathbf{U}^n) \quad \leftarrow \text{Relax by GS}$$

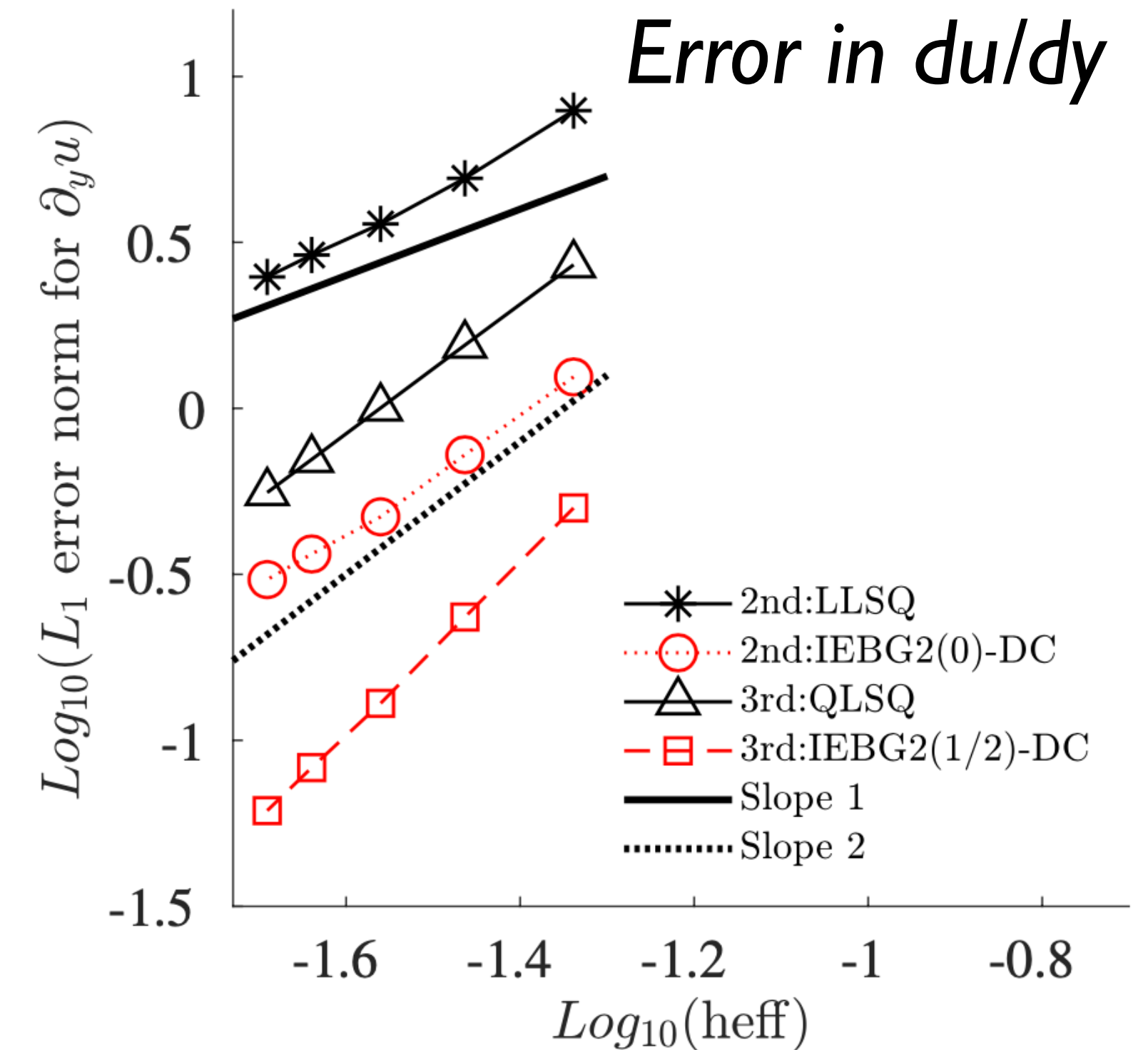
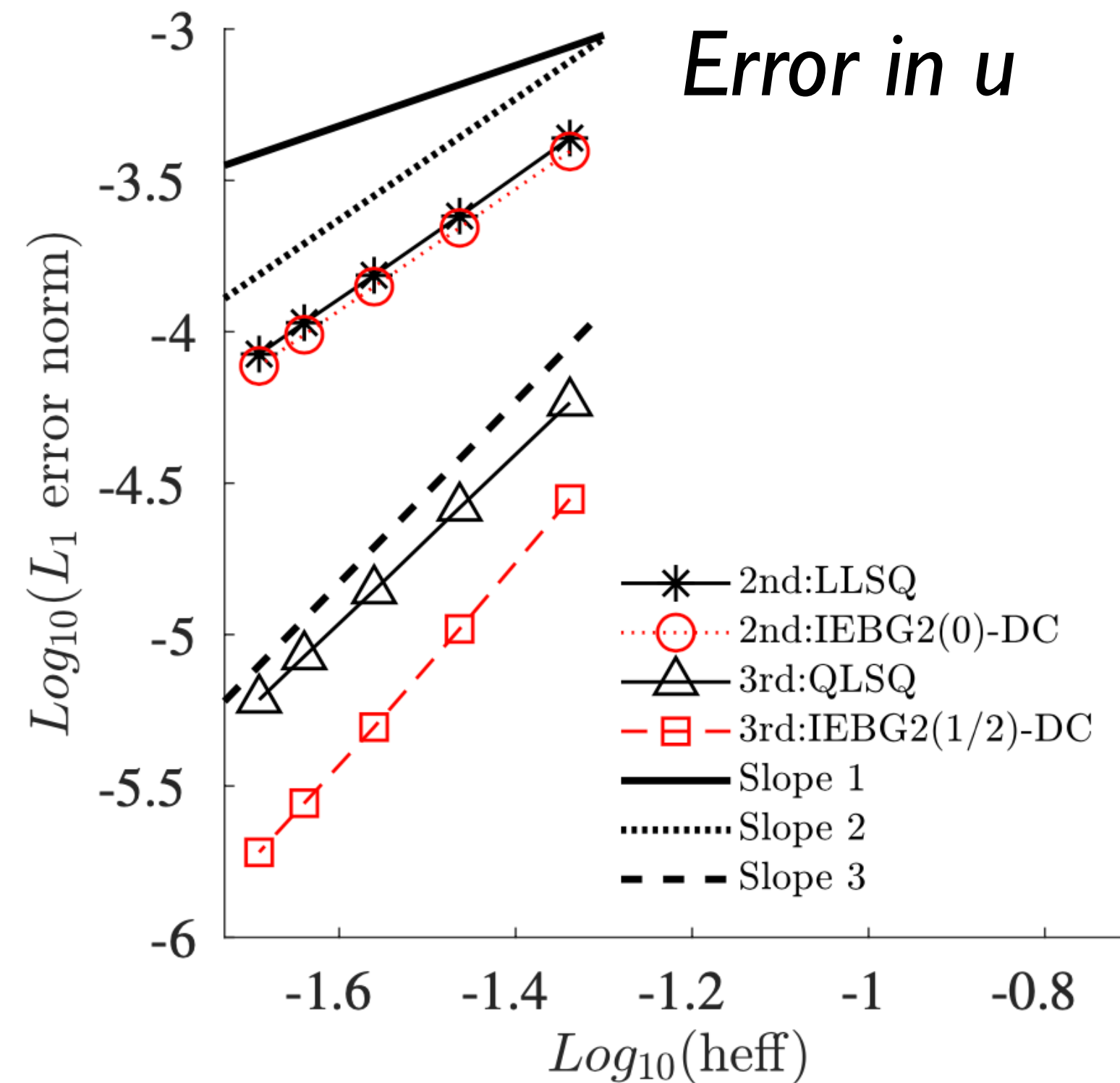
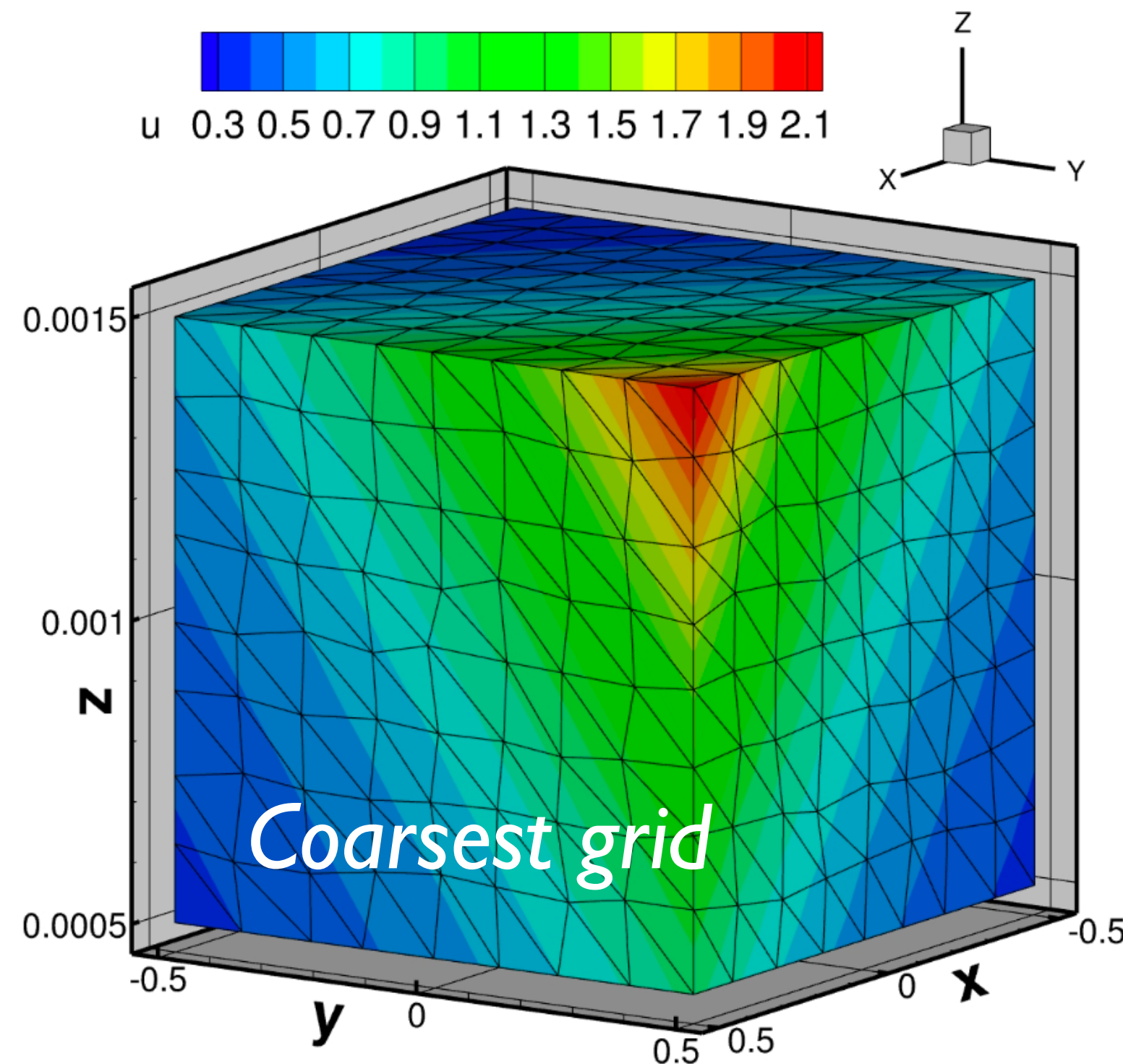
This is a loosely-coupled iteration between gradients and solutions.

Results

Accuracy Verification

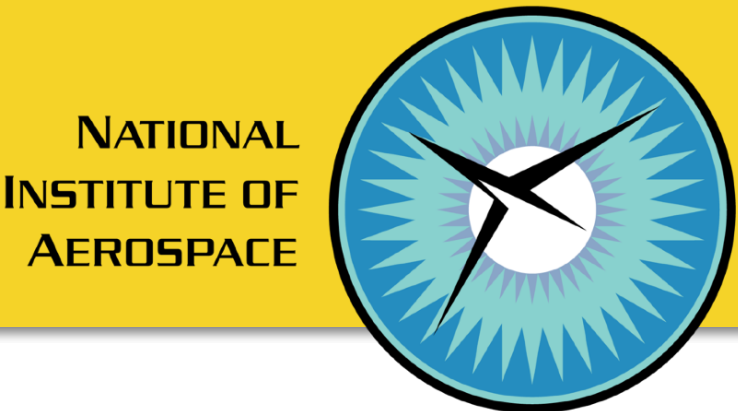


Solve Euler by 2nd- and 3rd-order NG-CCFV methods with LSQ and IEBG2 gradients over 5 levels of grids and **measure the solution error, using manufactured solutions.**



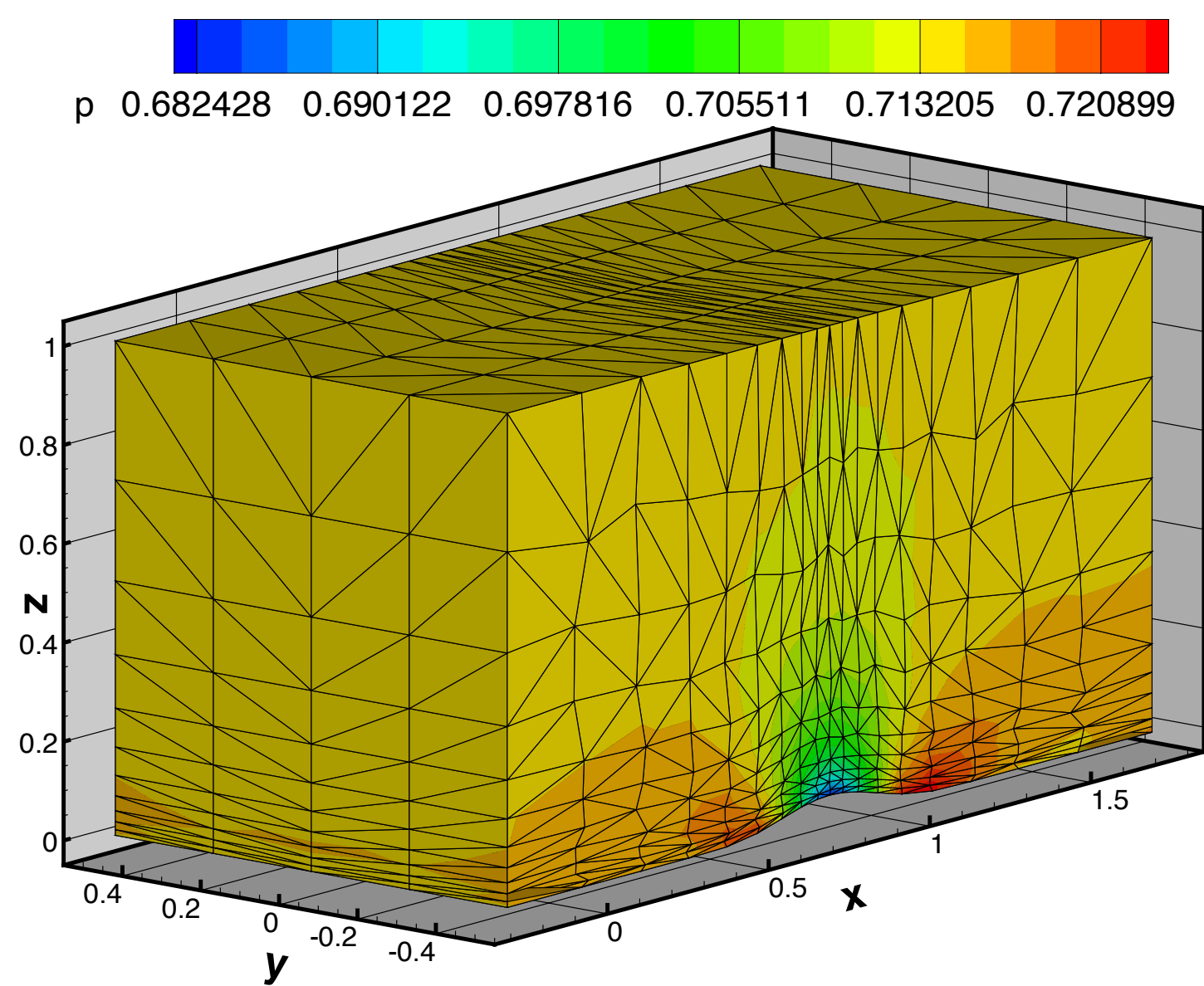
The solver took twice as many iteration with IEBG2. Not expected but...

Mach 0.3 Subsonic Flow over a smooth bump

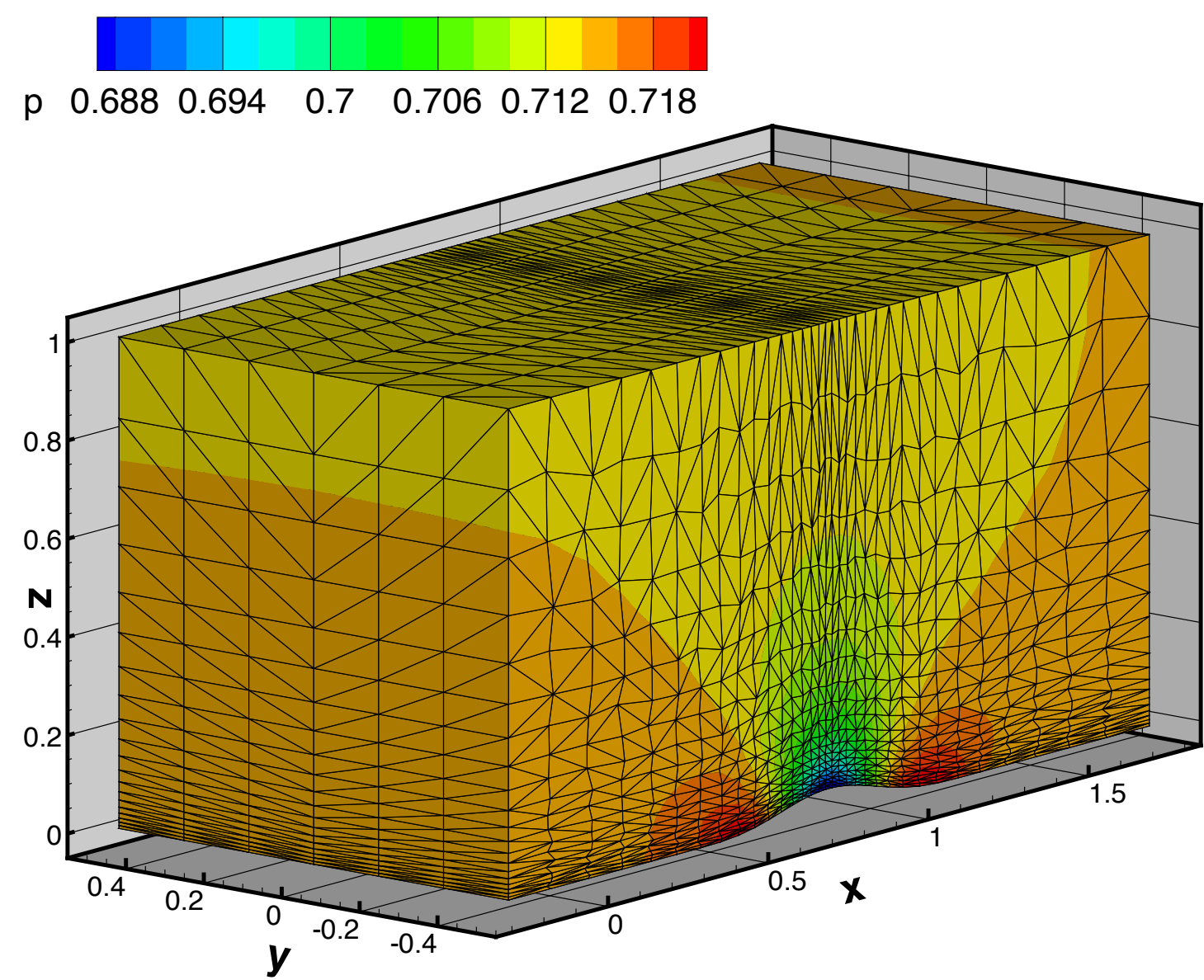


Solve Euler by 2nd- and 3rd-order NG-CCFV methods with LSQ and IEBG2 gradients over 5 levels of grids and **measure the spurious entropy error.**

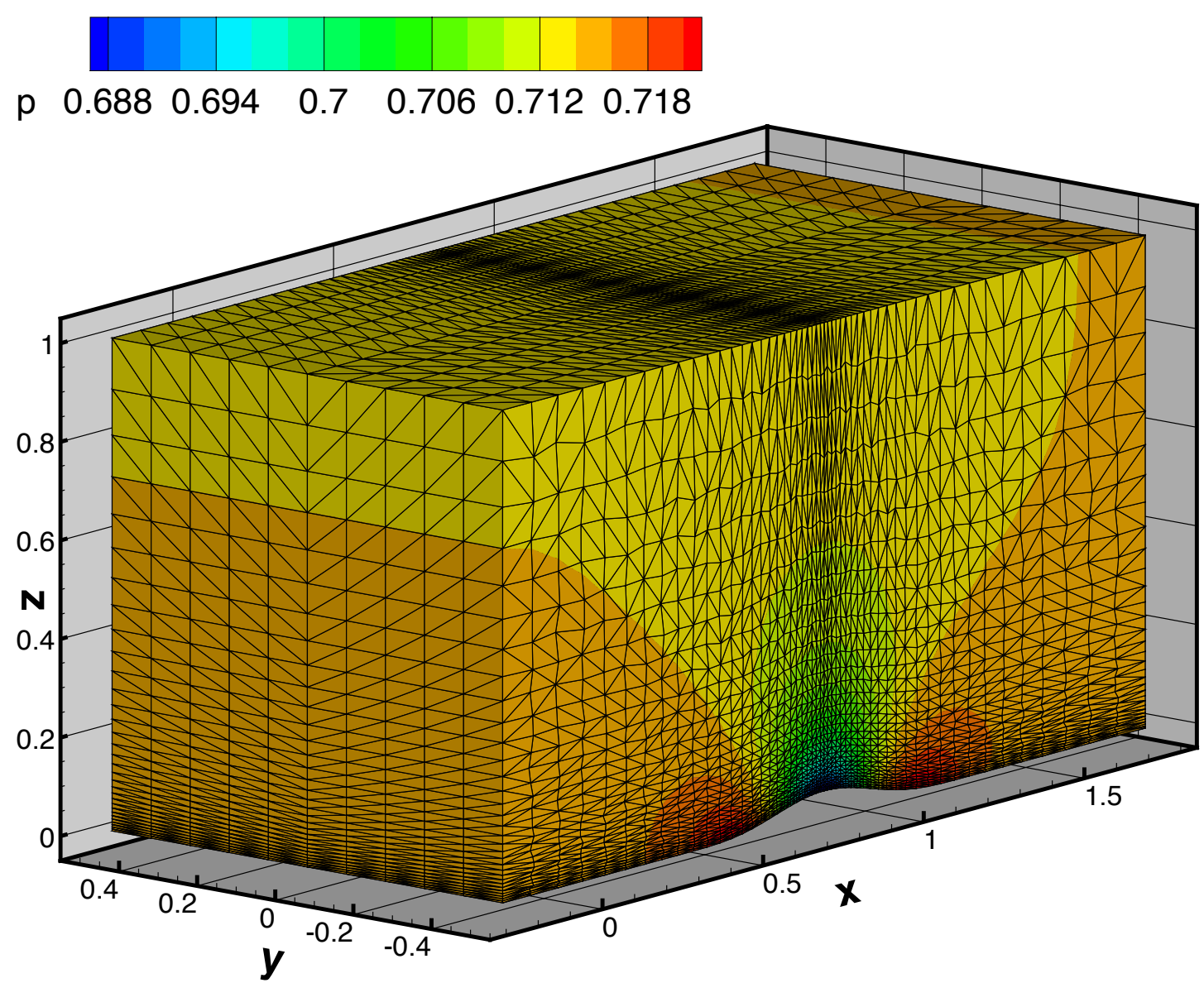
Grid 1



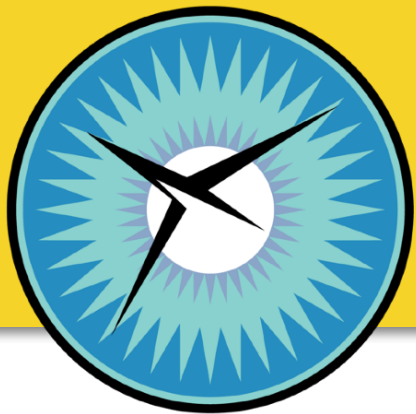
Grid 2



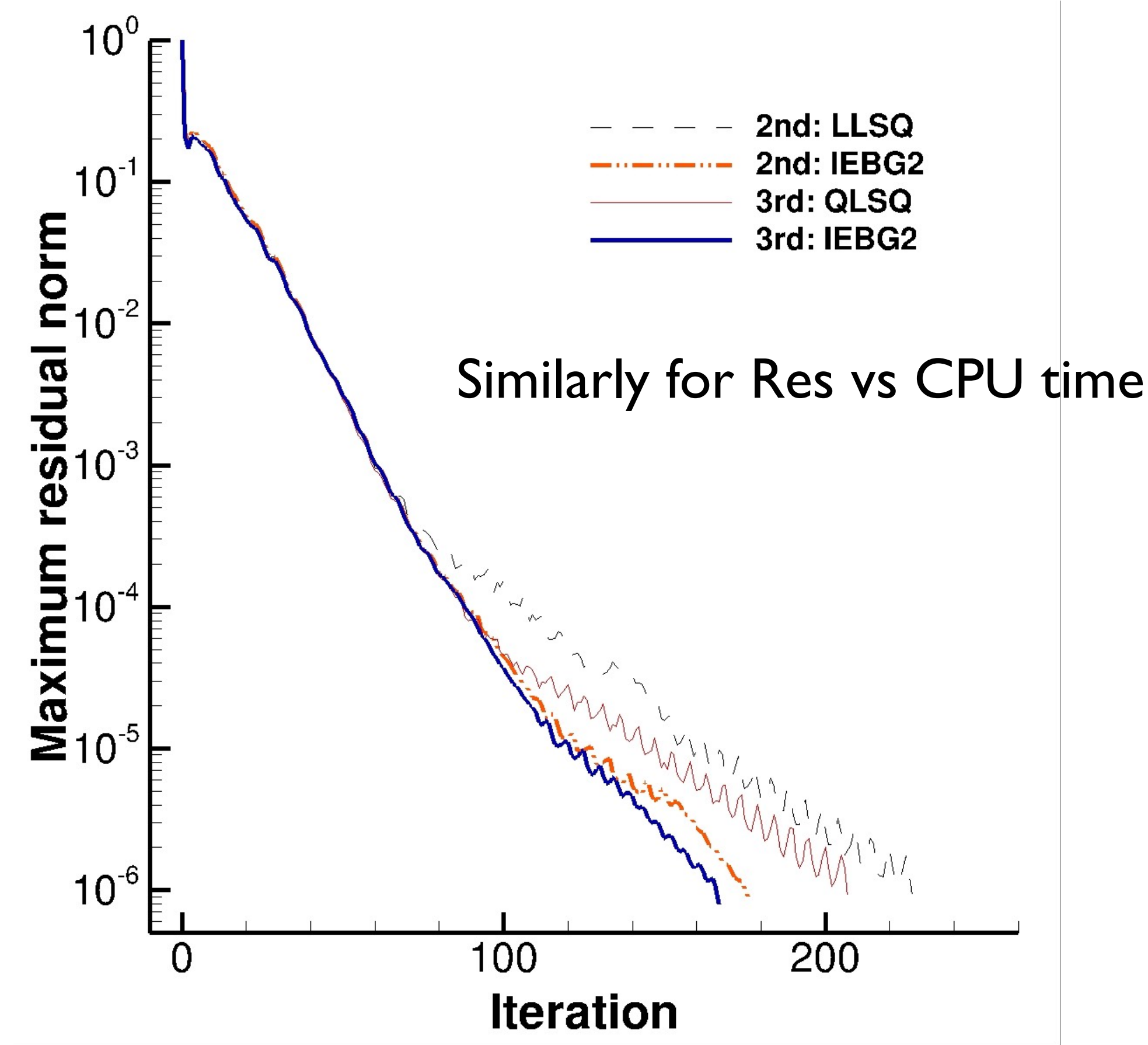
Grid 3



Subsonic Flow: $\kappa = 0$ and $c = 5$

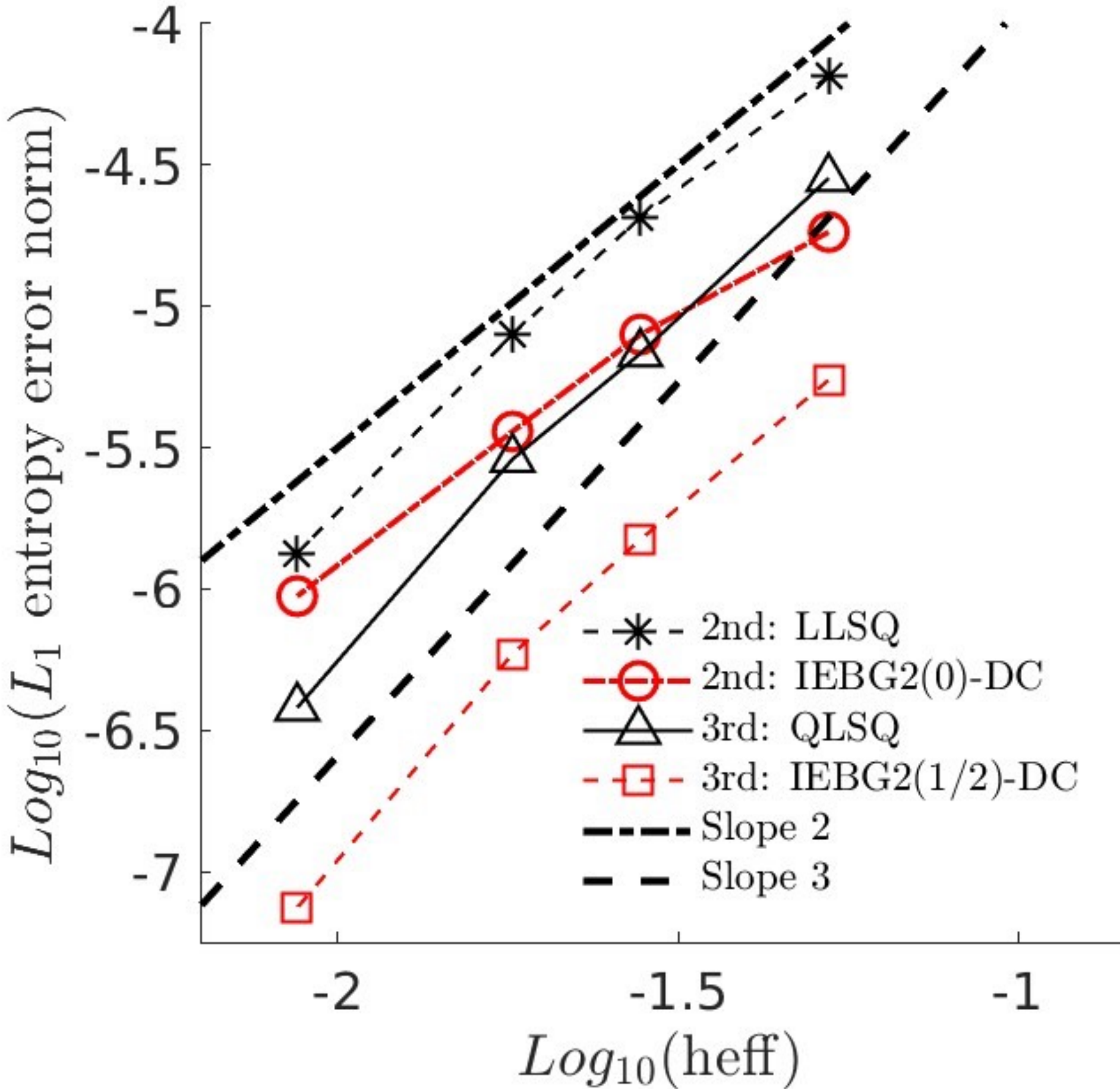


(1) Residual norm vs iteration

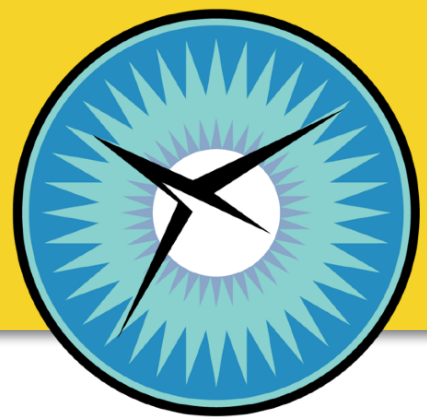


Fewer iterations and faster time-to-solution
IEBG methods are not expensive.

(2) Entropy error convergence



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



Conclusions:

- Developed simple implicit gradient methods for 2nd/3rd-order NG-CCFV methods.
- 3rd-order NG-CCFV method remains 2nd-derivative-free.
- Significantly lower gradient errors by implicit gradient methods.

Future work:

- Compact-stencil version depending only on solutions at cells around a node.
- JFNK solver with gradients added as additional unknowns.
- Further tests needed for more realistic problems with shocks.