

Comparison of Continuous and Discontinuous Galerkin methods in incompressible fluid flow problems

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Overview of CG and DG

Continuous Galerkin (CG)

- ✓ Well established
- ✓ Computationally efficient
- ✓ Static condensation
- ✗ Unstable for high convective flows
- ✗ Cumbersome p-adaptivity

DG (DG, IPM, CDG, LDG..)

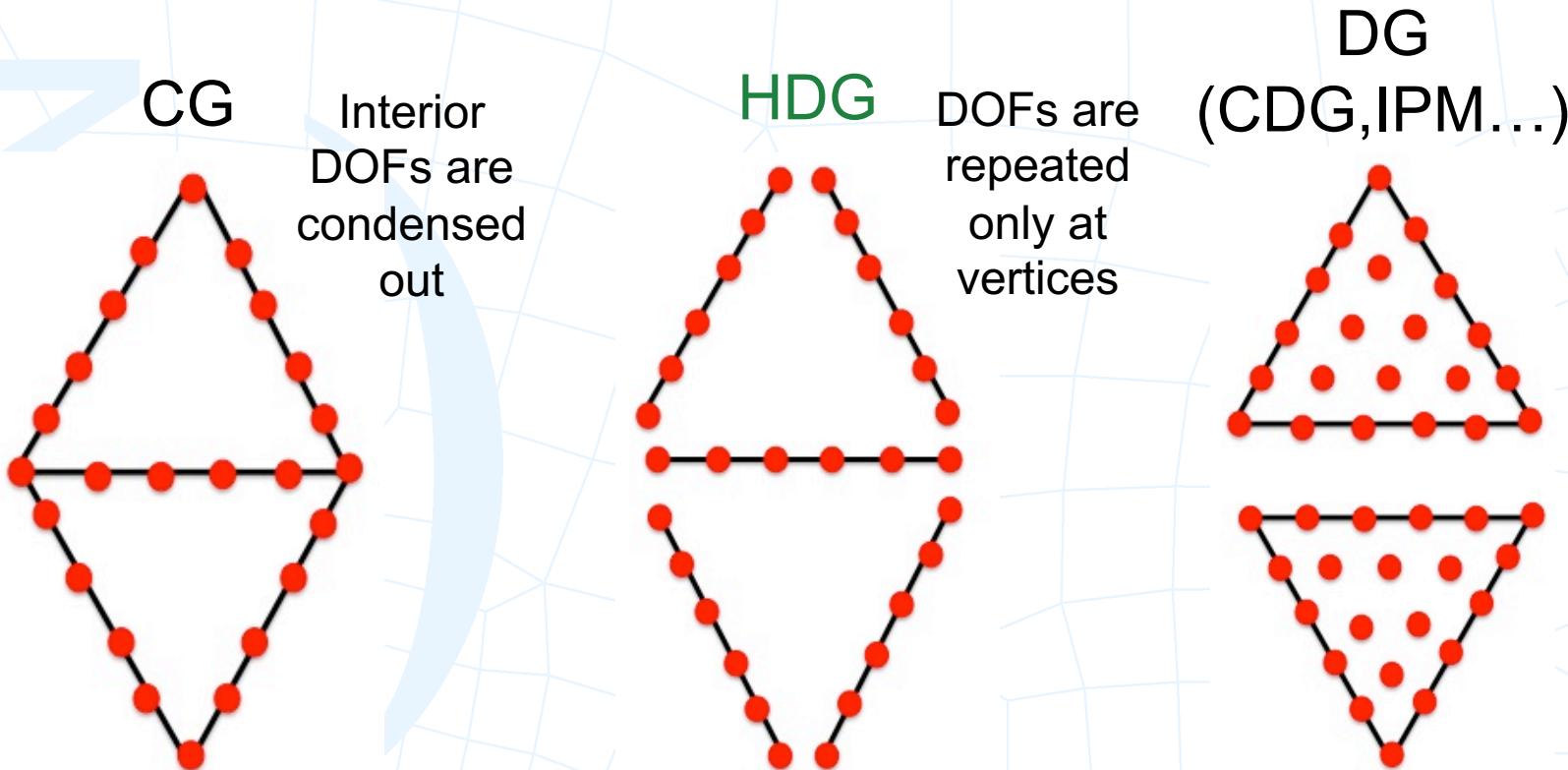
- ✗ Duplication of nodes
- ✗ No static condensation
- ✓ Stable for convective flows
- ✓ Suitability for adaptivity, parallel computing,...

Hybridizable Discontinuous
Galerkin (HDG)

Computationally
efficient?

Sailent features of HDG

- Less DOFs compared to other DG methods

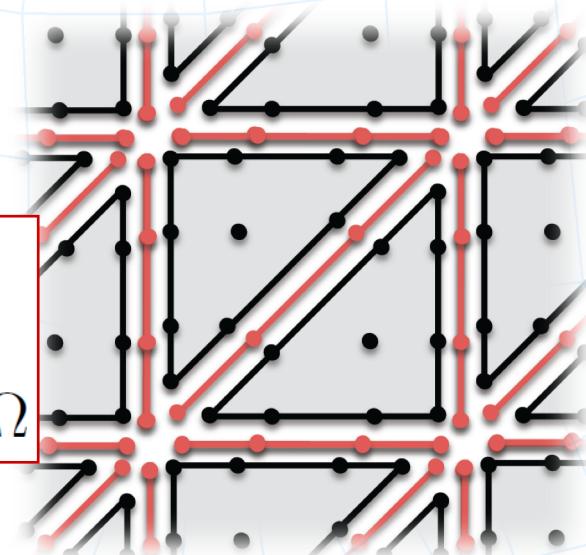


- Convergence of order $k+1$ for all variables: velocity, 1st derivatives, and pressure
- Superconvergence (order $k+2$ for velocity) with local postprocessing

1. Introduction to HDG

Incompressible Navier-Stokes equations

$$\begin{aligned} \nabla \cdot (\mathbf{u} \otimes \mathbf{u}) - \nabla \cdot (-p\mathbf{I} + \nu \nabla \mathbf{u}) &= f \quad \text{in } \Omega \\ \nabla \cdot \mathbf{u} &= 0 \quad \text{in } \Omega \\ \mathbf{u} &= \mathbf{u}_D \quad \text{on } \partial\Omega \end{aligned}$$



Consider an element K_i in FE mesh

Equivalent problem inside each element

$$\Gamma := \bigcup_{i=1}^{n_{\text{el}}} \partial K_i$$

Global Problem

Local Problem

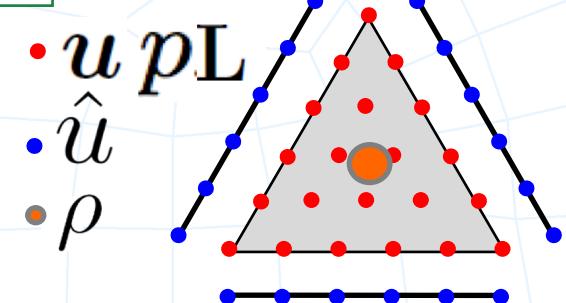
$$\begin{aligned} \mathbf{L} - \nabla \mathbf{u} &= 0 && \text{in } K_i \\ \nabla \cdot (\mathbf{u} \otimes \mathbf{u}) - \nabla \cdot (-p\mathbf{I} + \nu \mathbf{L}) &= f && \text{in } K_i \\ \nabla \cdot \mathbf{u} &= 0 && \text{in } K_i \\ \mathbf{u} = \hat{\mathbf{u}} & && \text{on } \partial K_i \\ \frac{1}{|K_i|} \int_{K_i} p \, dV &= \rho^{K_i} && \text{in } K_i \end{aligned}$$

$$\begin{aligned} [(-p\mathbf{I} + \nu \mathbf{L}) \cdot \mathbf{n}] &= 0 && \text{on } \Gamma \setminus \partial\Omega \\ \int_{\partial K_i} \hat{\mathbf{u}} \cdot \mathbf{n} \, dS &= 0 && \text{on } \Gamma \setminus \partial\Omega \\ \mathbf{u} &= \mathbf{u}_D && \text{on } \partial\Omega \end{aligned}$$

HDG Local Problem

$$\begin{aligned}
 \mathbf{L} - \nabla \mathbf{u} &= 0 \quad \text{in } K_i \\
 \nabla \cdot (\mathbf{u} \otimes \mathbf{u}) - \nabla \cdot (-p\mathbf{I} + \nu\mathbf{L}) &= \mathbf{f} \quad \text{in } K_i \\
 \nabla \cdot \mathbf{u} &= 0 \quad \text{in } K_i \\
 \mathbf{u} &= \hat{\mathbf{u}} \quad \text{on } \partial K_i \\
 \frac{1}{|K_i|} \int_{K_i} p \, dV &= \rho^{K_i} \quad \text{in } K_i
 \end{aligned}$$

HDG discretization leads to a non-linear system **in each element**. The Newton-Raphson linealization is: given $\delta\hat{\mathbf{u}}$, $\delta\rho$ find $\delta\mathbf{u}$, $\delta\mathbf{L}$ and $\delta\mathbf{p}$ s.t.



$$\begin{bmatrix} \mathbf{A}_{uu}(\mathbf{u}) & \mathbf{A}_{uL} & \mathbf{A}_{up} \\ \mathbf{A}_{Lu} & \mathbf{A}_{LL} & \\ \mathbf{A}_{pu} & & \end{bmatrix} \begin{bmatrix} \delta\mathbf{u} \\ \delta\mathbf{L} \\ \delta\mathbf{p} \end{bmatrix} = - \begin{bmatrix} \mathbf{r}_u \\ \mathbf{r}_L \\ \mathbf{r}_p \end{bmatrix} - \begin{bmatrix} \mathbf{A}_{u\hat{u}}(\hat{\mathbf{u}}) \\ \mathbf{A}_{L\hat{u}} \\ \mathbf{A}_{p\hat{u}} \end{bmatrix} \begin{bmatrix} \delta\hat{\mathbf{u}}_{F1} \\ \delta\hat{\mathbf{u}}_{F2} \\ \delta\hat{\mathbf{u}}_{F3} \end{bmatrix}$$

with the constraint $\mathbf{A}_{\rho p}\delta\mathbf{p} = \rho^k$

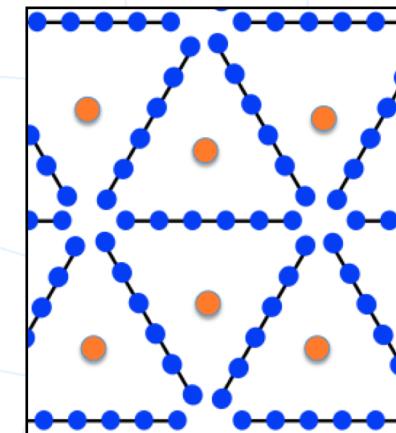
Local Solver

HDG Global problem

$$\begin{aligned} \llbracket (-p\mathbf{I} + \nu\mathbf{L}) \cdot \mathbf{n} \rrbracket &= 0 \quad \text{on } \Gamma \setminus \partial\Omega \\ \int_{\partial K_i} \hat{\mathbf{u}} \cdot \mathbf{n} \, dS &= 0 \quad \text{on } \Gamma \setminus \partial\Omega \\ \mathbf{u} &= \mathbf{u}_D \quad \text{on } \partial\Omega \end{aligned}$$



HDG discretization
NR linearization
replace local problem



- $\hat{\mathbf{u}}$: velocity trace
- ρ : mean pressure

$$\begin{bmatrix} \mathbf{A}_{\hat{\mathbf{u}}\hat{\mathbf{u}}}(\hat{\mathbf{u}}) & \mathbf{A}_{\rho\hat{\mathbf{u}}}^T \\ \mathbf{A}_{\rho\hat{\mathbf{u}}} & \end{bmatrix} \begin{bmatrix} \delta\hat{\mathbf{u}} \\ \delta\rho \end{bmatrix} = \begin{bmatrix} \mathbf{s} \\ \mathbf{0} \end{bmatrix}$$

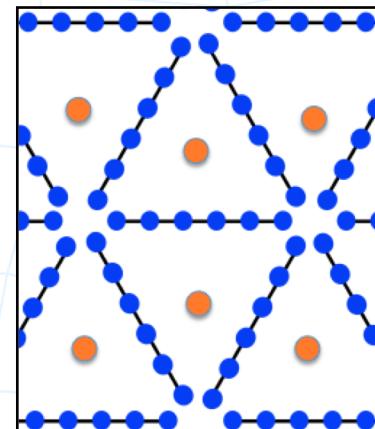
Global Solver

computations

1. Global problem

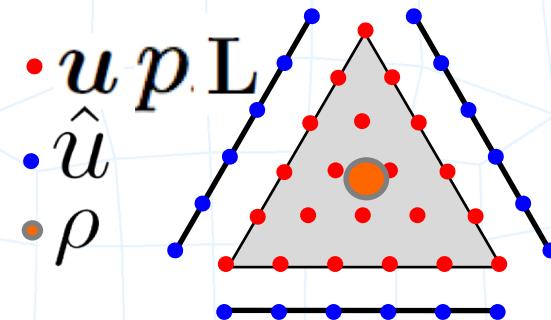
$$\begin{bmatrix} \mathbf{A}_{\hat{u}\hat{u}}(\hat{\mathbf{u}}) & \mathbf{A}_{\rho\hat{u}}^T \\ \mathbf{A}_{\rho\hat{u}} & \end{bmatrix} \begin{bmatrix} \delta\hat{\mathbf{u}} \\ \delta\rho \end{bmatrix} = \begin{bmatrix} \mathbf{s} \\ \mathbf{0} \end{bmatrix}$$

*matrix with nice
block structure*



2. Cheap element-by-element postprocess (local solver)

$$\delta\hat{\mathbf{u}} \ \delta\rho \rightarrow \delta\mathbf{u}^K \delta\mathbf{L}^K \delta\mathbf{p}^K$$



3. Cheap element-by-element 2nd postprocess → superconvergent solution (order k+2 for u)

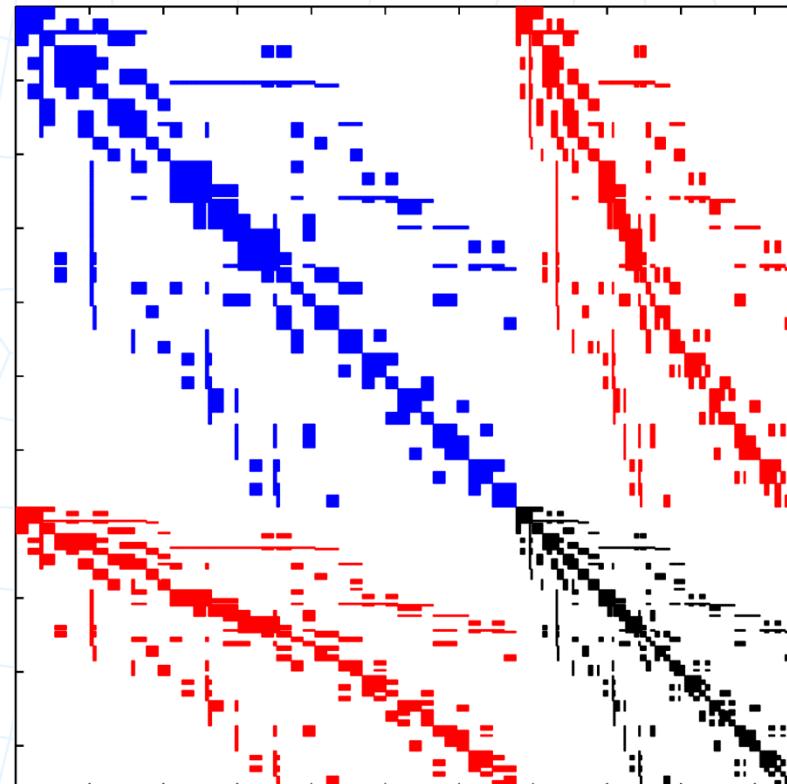
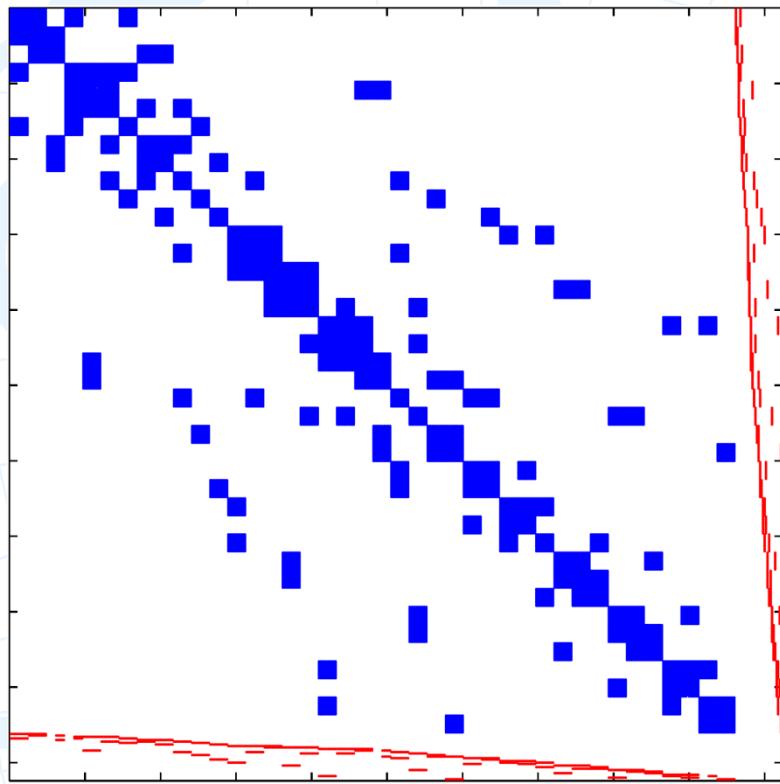
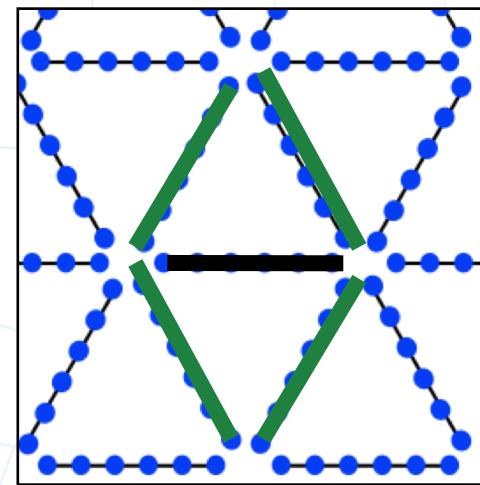
HDG block structure

- Each face has contributions from 4 faces of neighbouring elements.
- Each row has **5 blocks** of same size

HDG is more efficient for linear solvers

HDG #ndofs = 511

CG #ndofs = 530



Degree
k = 5

HDG super convergence

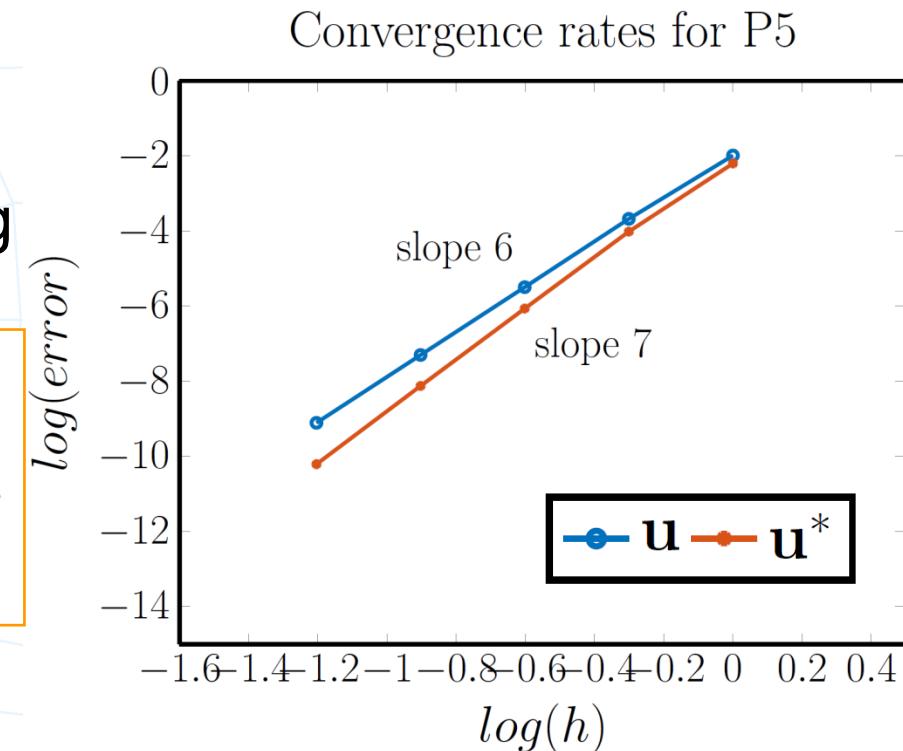
- Optimal convergence: **order $k+1$** for u , its **gradient**, and p
- Element-by-element postprocess** in $P^{k+1}(K)$ solving the following equations

$$\begin{aligned} \nabla \cdot \nabla u^* &= \nabla \cdot L \quad \text{in } K_i \\ n \cdot \nabla u^* &= n \cdot L \quad \text{on } \partial K_i \\ \int_{K_i} u^* dV &= \int_{K_i} u dV \end{aligned}$$

leads to a superconvergent solution u^* , with convergence of **order $k+2$**



Inexpensive and reliable
Error Estimator



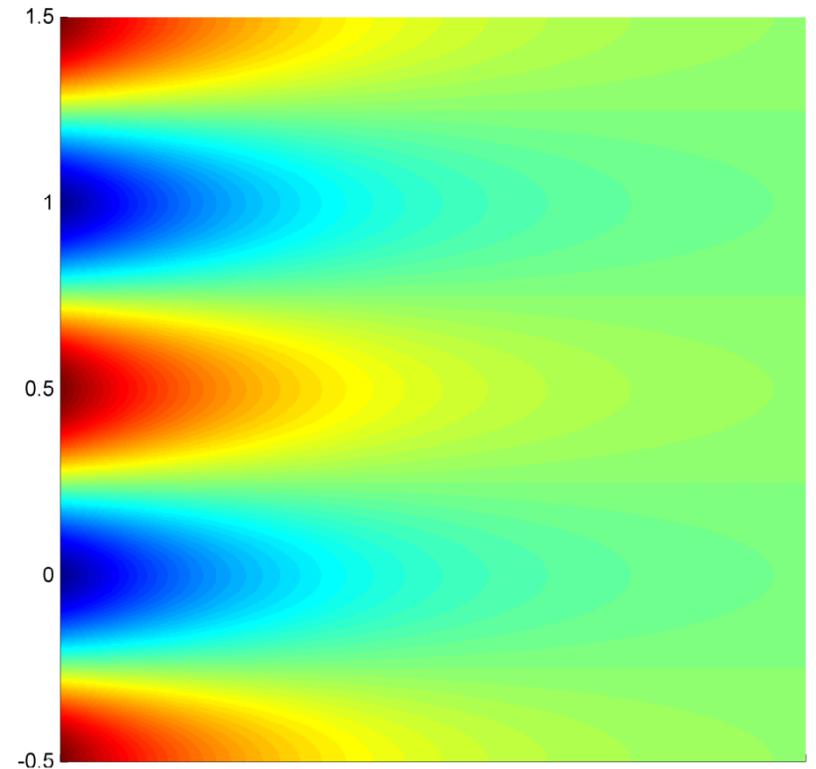
$$u - u^*$$

EFFICIENCY COMPARISON CG vs HDG

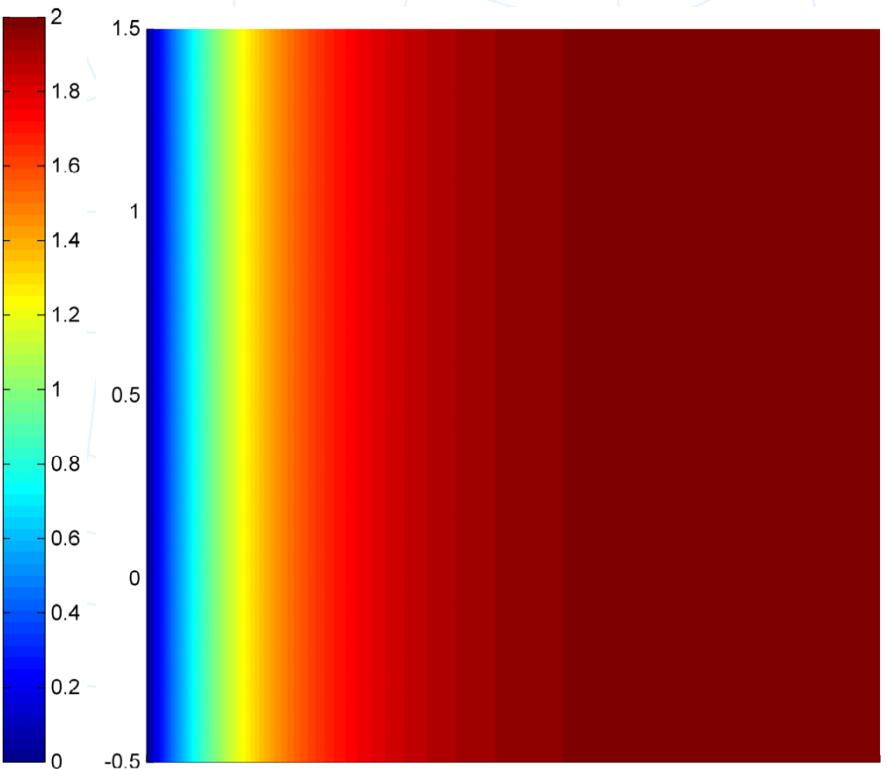
Numerical examples

Analytical solution: Kovasznay flow at $Re = 20$ in

$[0 \ 2] \times [-0.5 \ 1.5]$

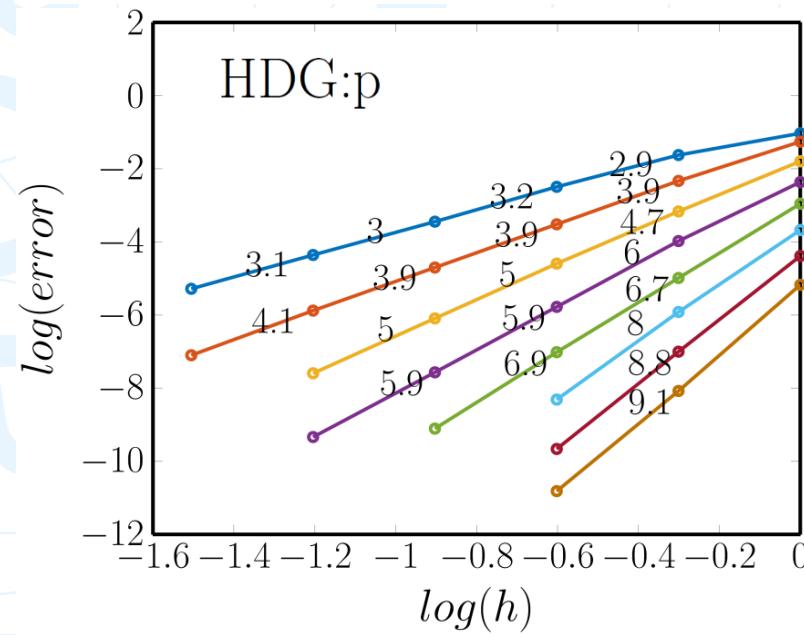
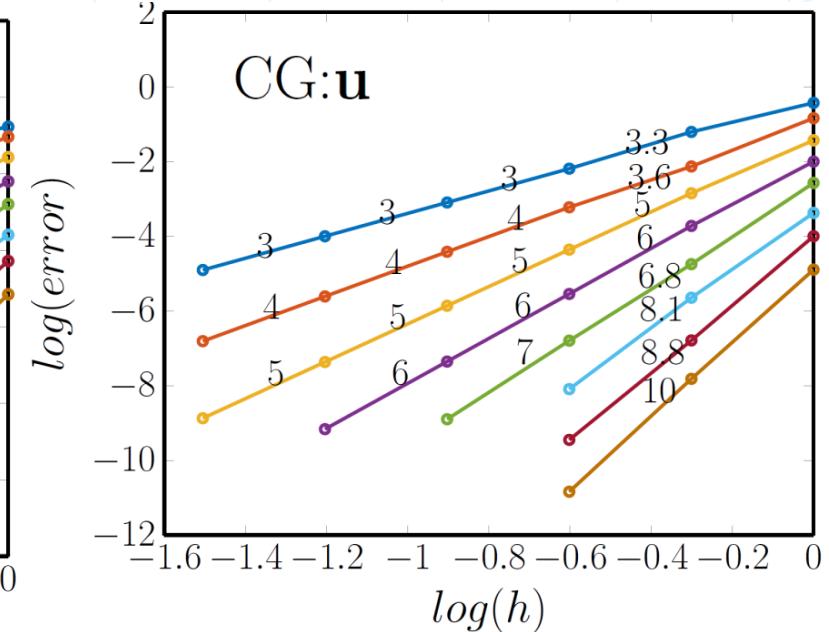
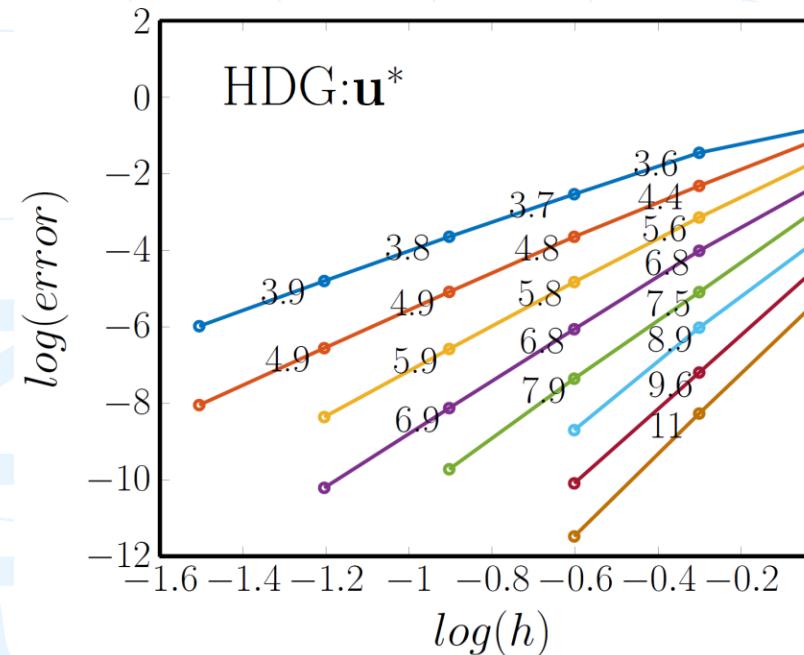


Velocity



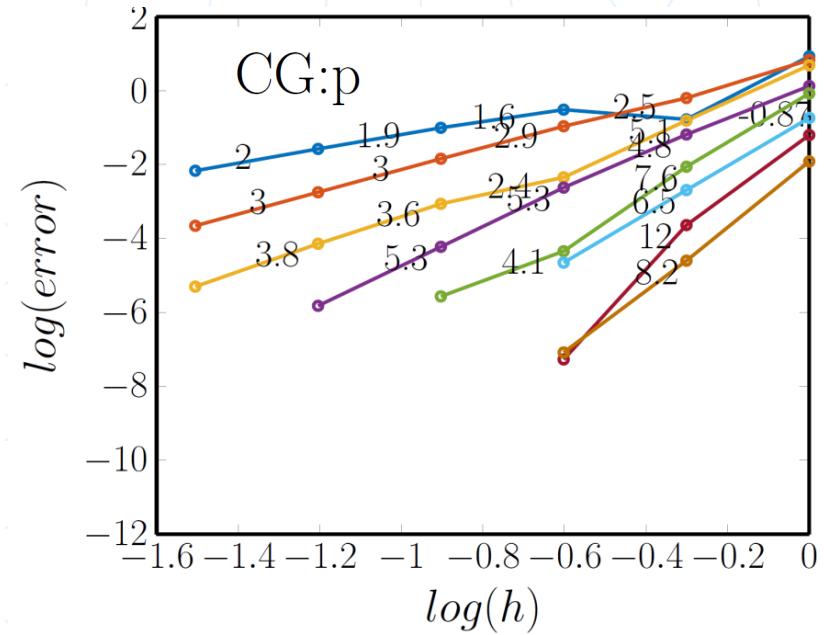
Pressure

Stokes: Convergence Plots



Legend:

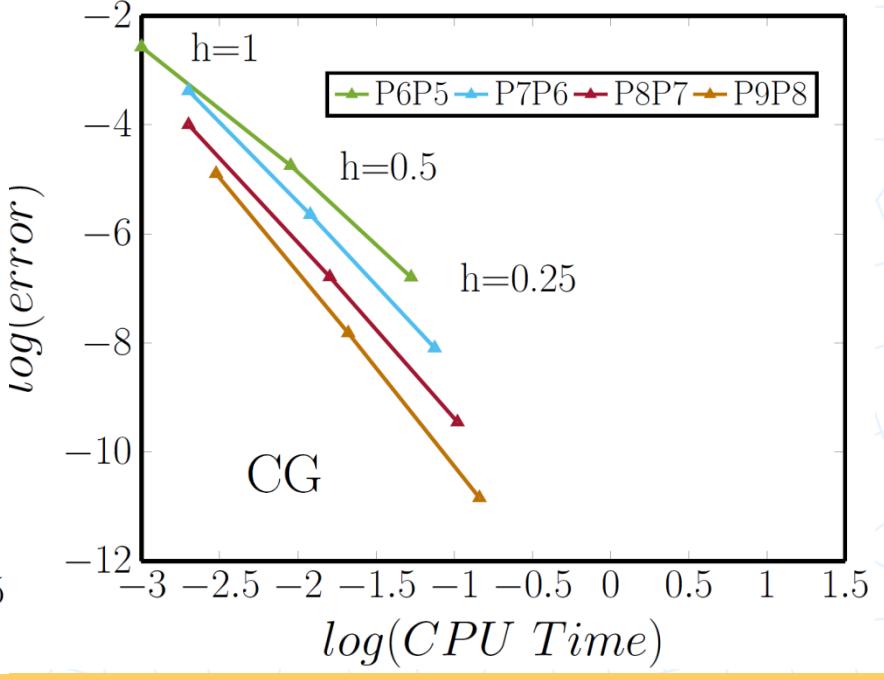
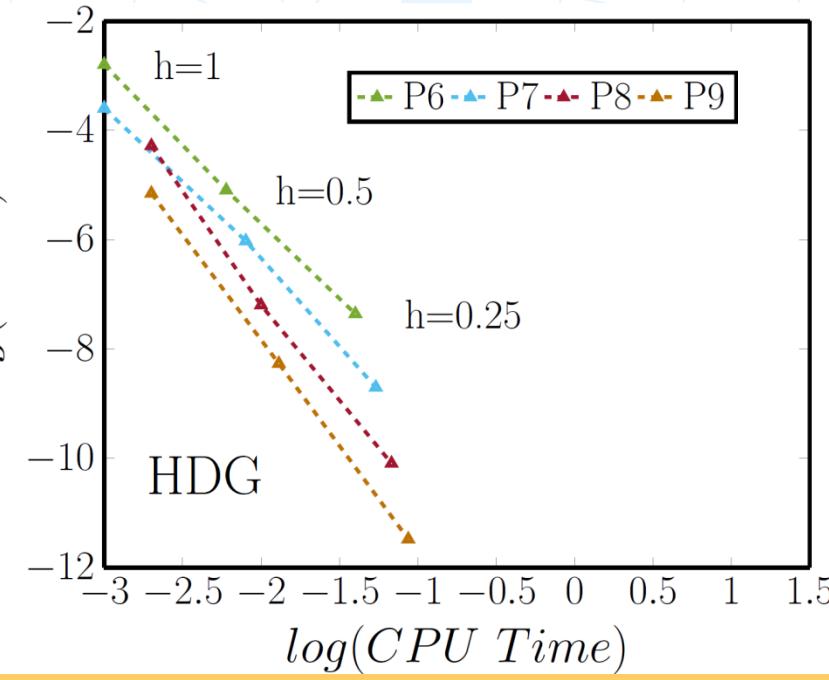
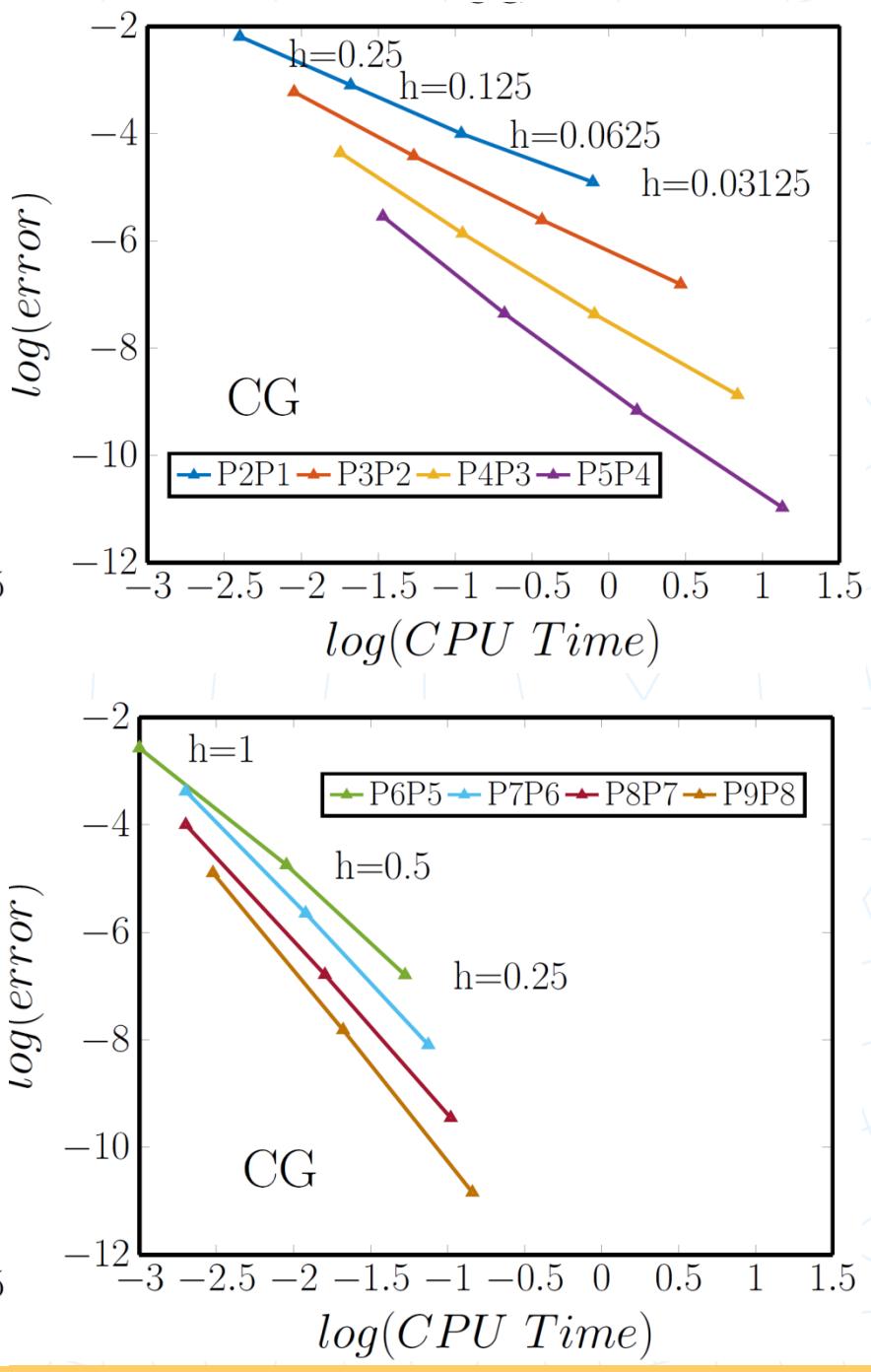
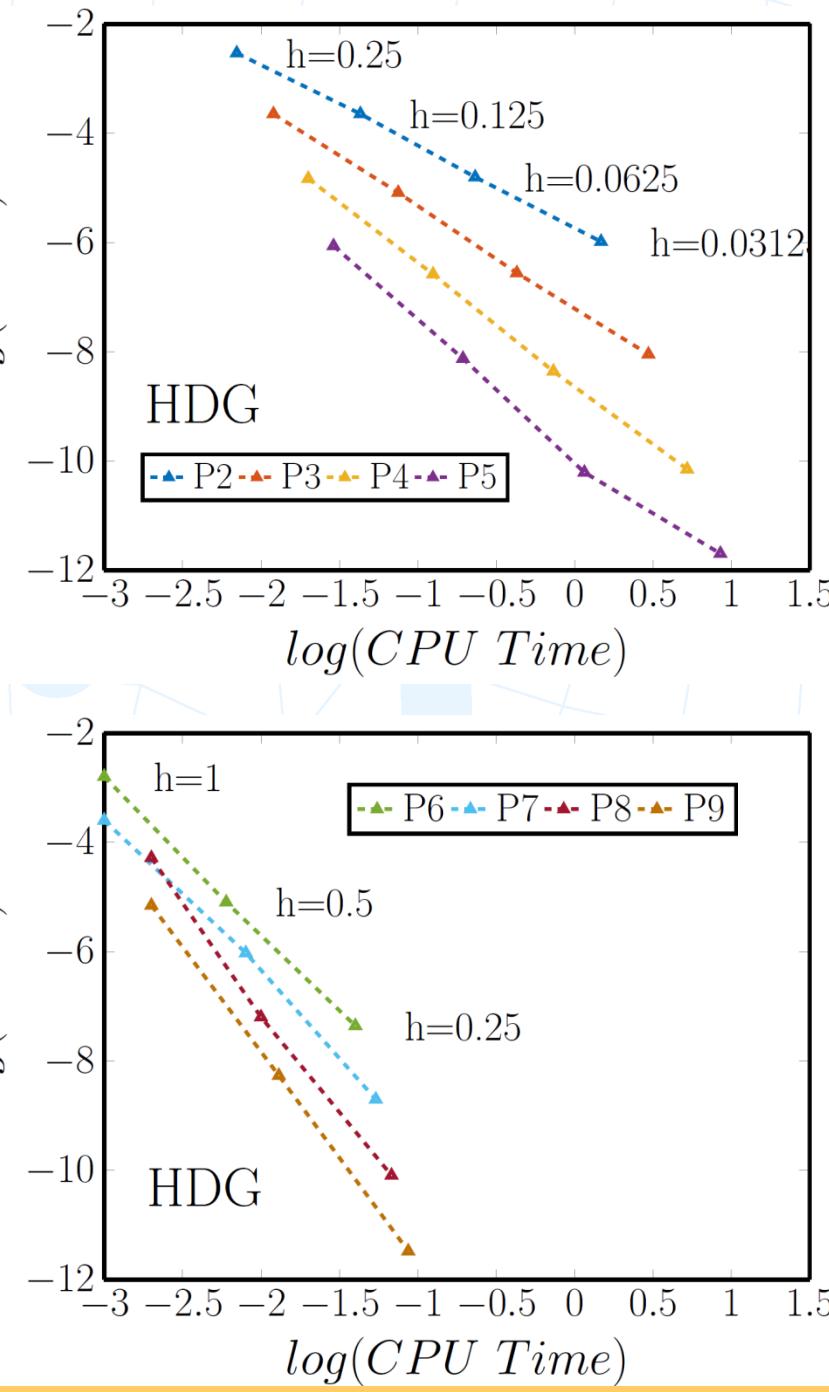
- P2P1 —●— P3P2 —●— P4P3 —●— P5P4
- P6P5 —●— P7P6 —●— P8P7 —●— P9P8



Legend:

- P2P1 —●— P3P2 —●— P4P3 —●— P5P4
- P6P5 —●— P7P6 —●— P8P7 —●— P9P8

Stokes: CPU Time for linear solver

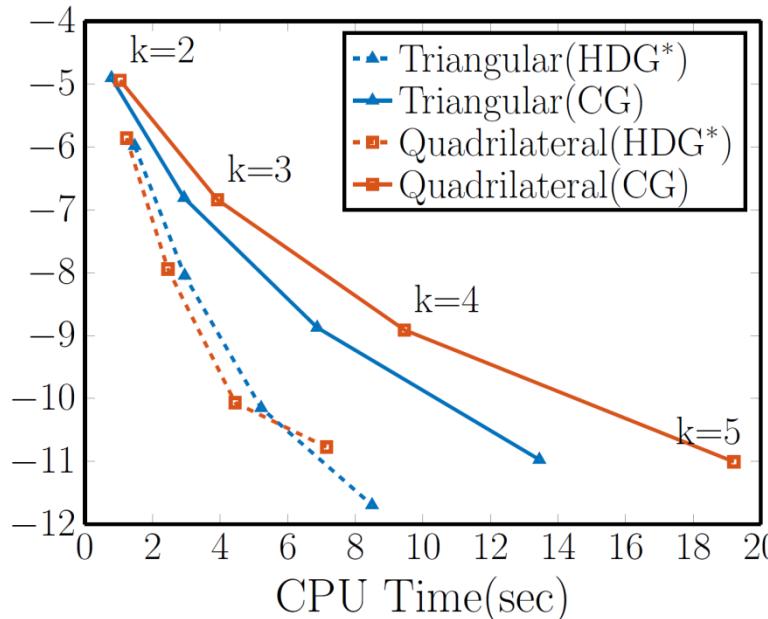


Stokes: CPU Time for linear solver

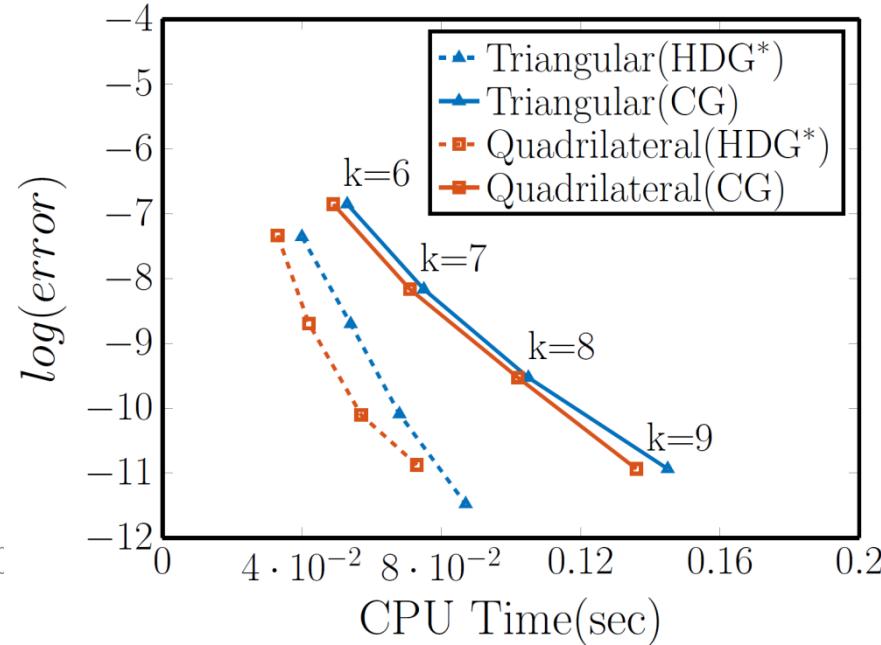
Similar trend is observed in the case of Navier-Stokes.

$$h = 0,03125$$

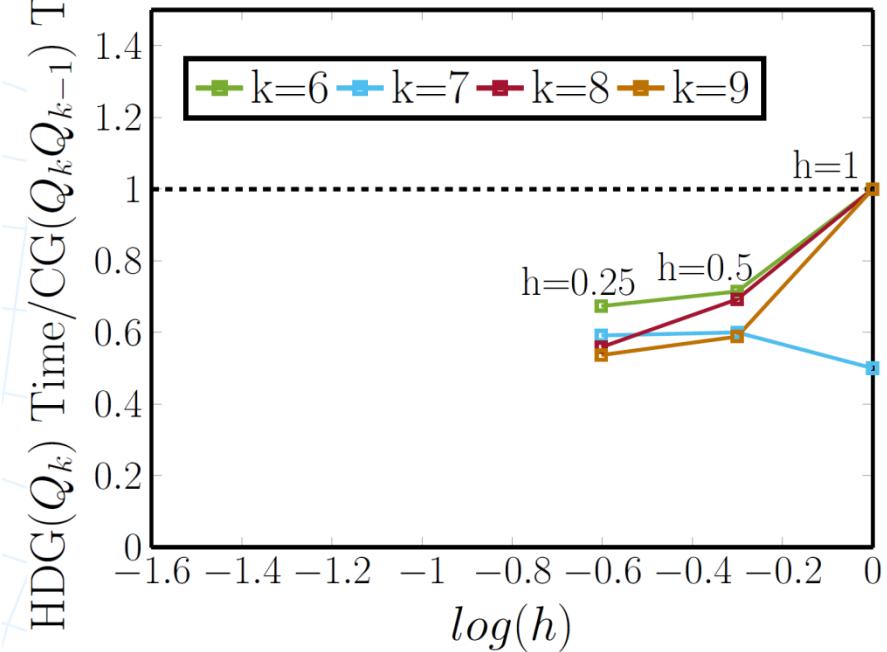
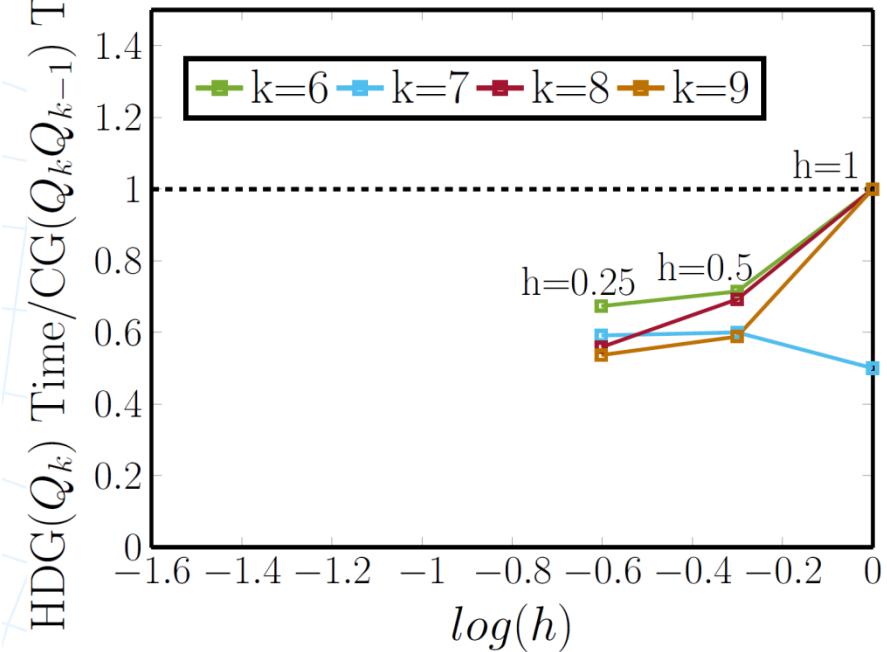
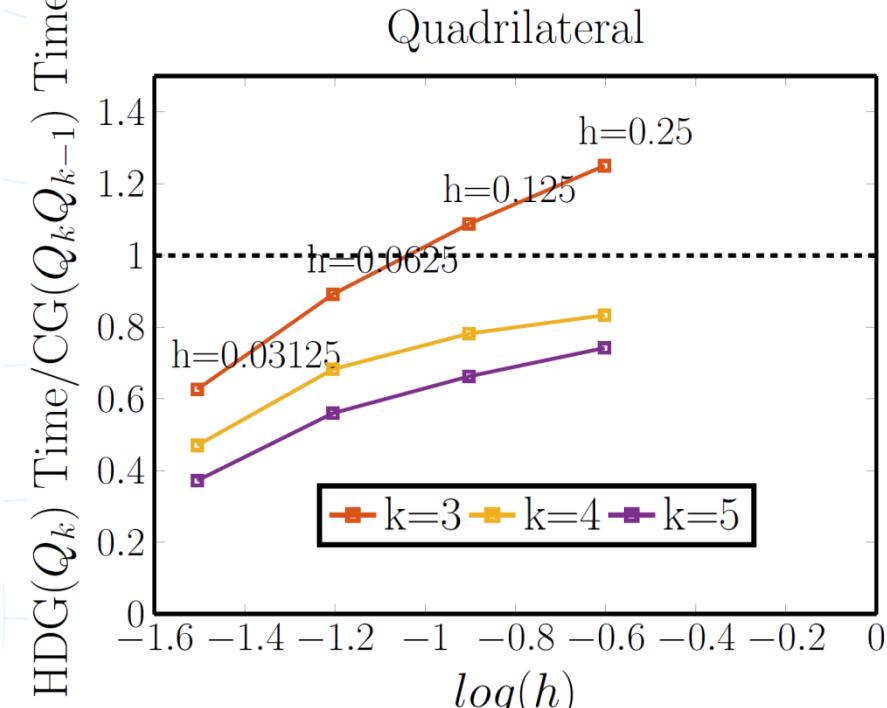
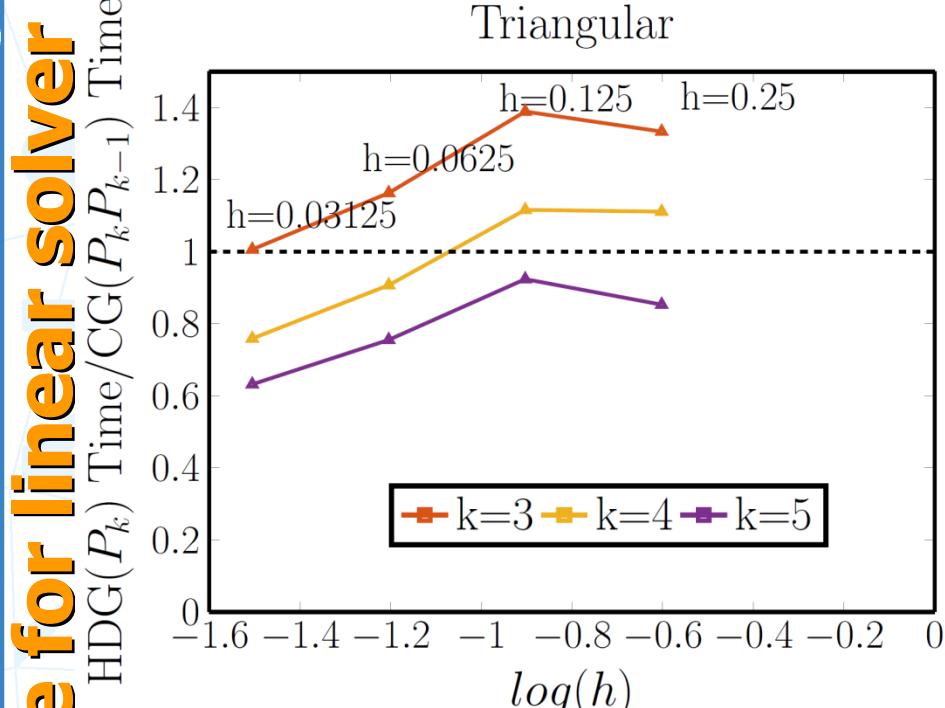
HDG: P_k & CG: $P_k P_{k-1}$



HDG: P_k & CG: $P_k P_{k-1}$



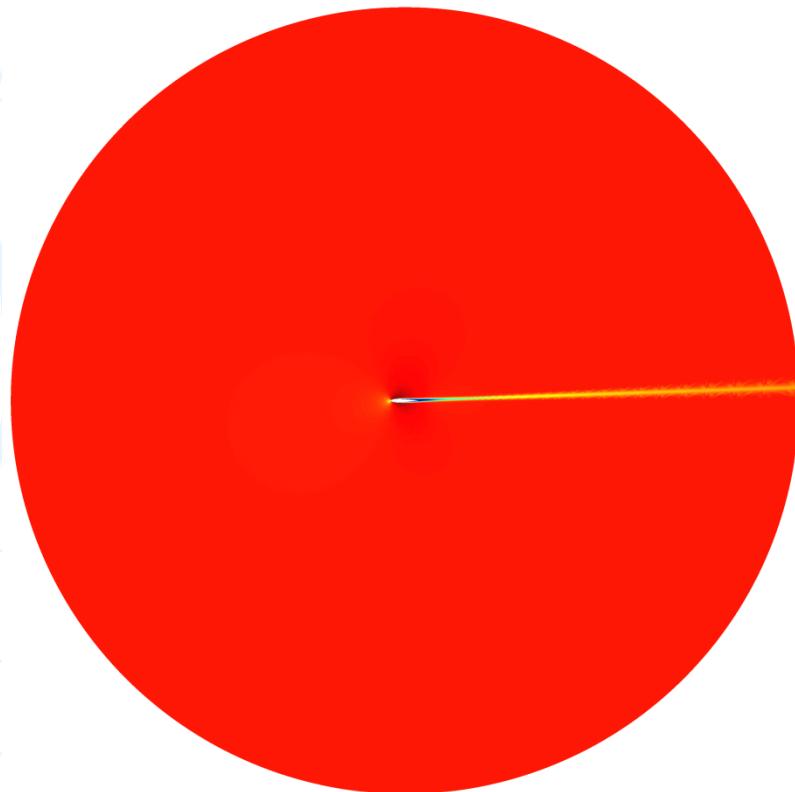
$$h = 0,25$$



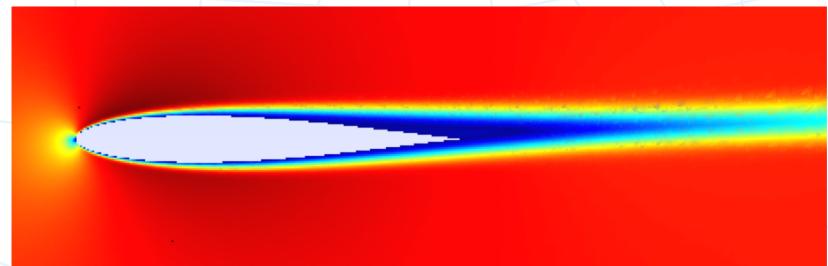
Stokes: CPU Time for linear solver

NACA0012 Airfoil

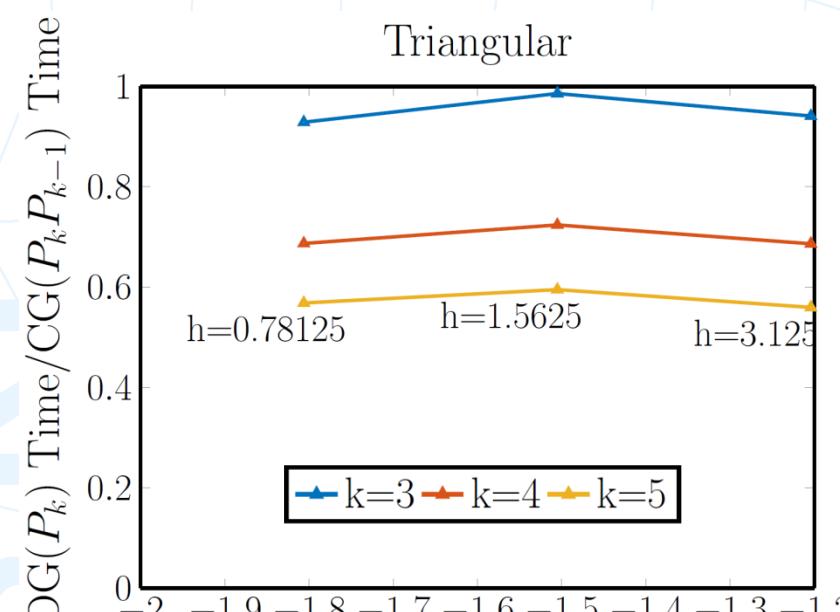
Incompressible Navier-Stokes is solved at $Re = 5000$ and Angle of attack (AoA) = 2° ; Error in lift coefficient is compared



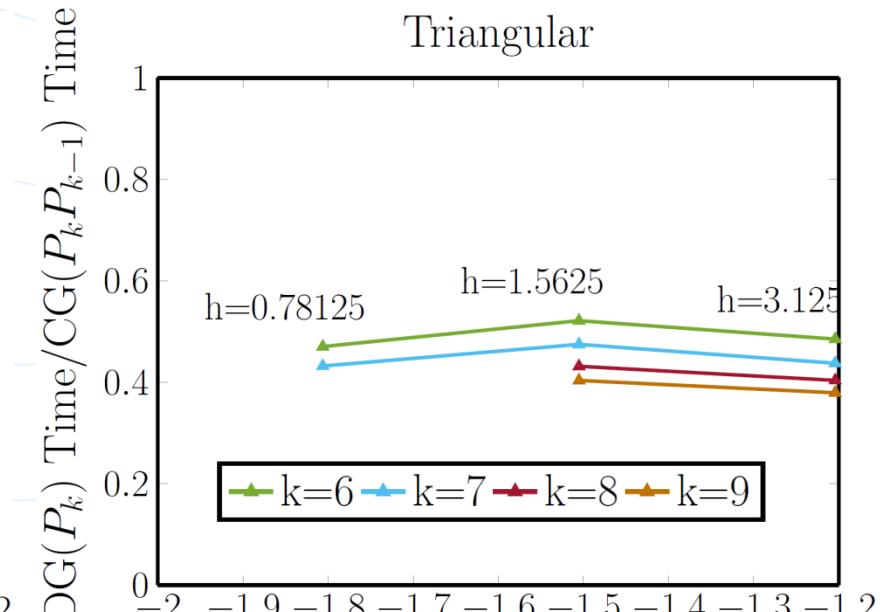
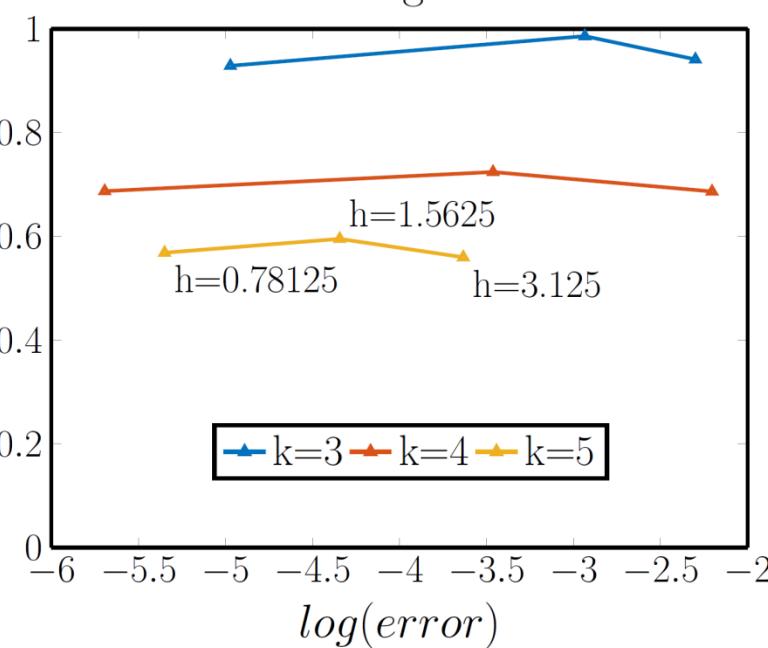
Velocity



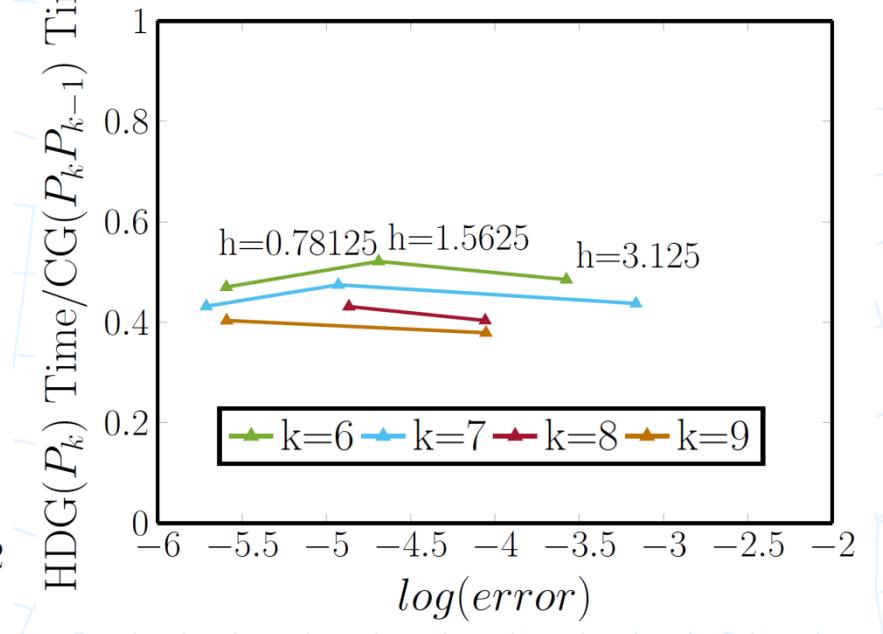
NACA: CPU Time for linear solver



Triangular



Triangular



Concluding remarks

- Computational efficiency between HDG and CG for Stokes problem
 - High order-coarser mesh is more efficient than low order-finer mesh in both HDG and CG.
 - HDG is more efficient at order $k \geq 3$ for same level of accuracy, in terms of CPU time of linear solver
- Similar computational efficiency is observed in Navier-Stokes example (NACA).