

Development and Validation of a Density-Based Implicit Solver Using LU-SGS Algorithm

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1. Background

2. Implicit Finite Volume Discretization

3. LU-SGS Algorithm

4. Results

5. Concluding Remarks

Outline

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Background

▶ Based on *DensityBasedTurbo* by Oliver Borm

- Density Based Coupled Algorithm
- Explicit Time Integration
- Godunov Type Flux Schemes
- Multi-Dimensional Slope Limiter
- Local Time Stepping
- Steady & Transient Solvers

▶ We have focused on steady state solver

- Implementation of implicit time integration
- Implementation of far-field boundary condition
 - Utilizing riemann invariants

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Implicit Finite Volume Discretization

► Favre-Averaged Navier-Stokes Equations in Integral Form

$$\int_V \frac{\partial \vec{W}}{\partial t} dV + \oint_S (\vec{F}_c - \vec{F}_v) dS = 0$$

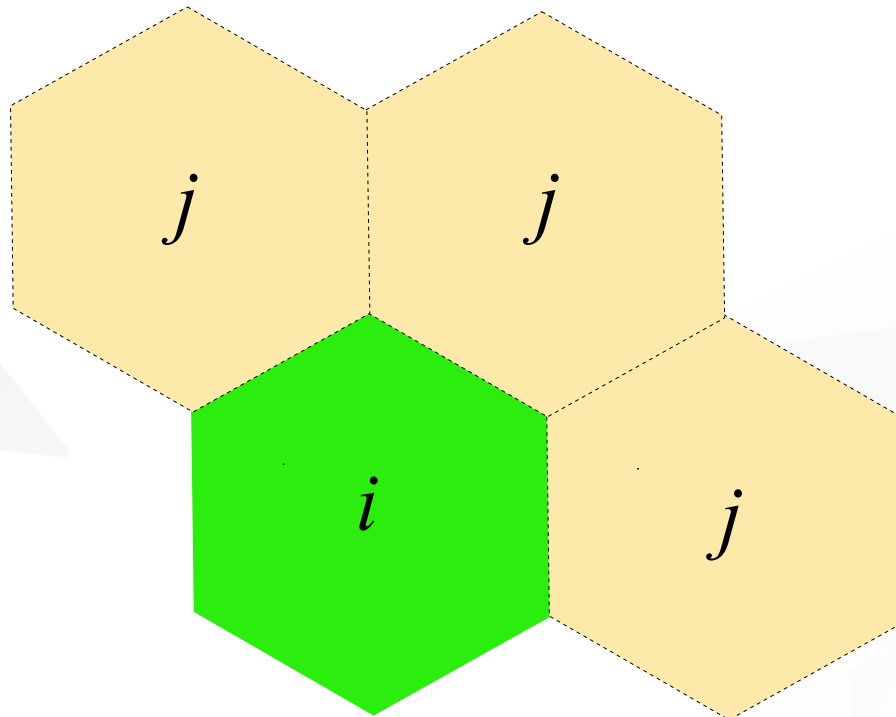
$$\vec{W} = \begin{bmatrix} \rho \\ \rho \vec{U} \\ \rho E \end{bmatrix} \quad \vec{F}_c = \begin{bmatrix} (\rho \vec{U})_f \cdot \vec{n} \\ (\rho \vec{U} \otimes \vec{U} + p \vec{I})_f \cdot \vec{n} \\ (\rho H \vec{U})_f \cdot \vec{n} \end{bmatrix}$$

$$\vec{F}_v = \begin{bmatrix} 0 \\ \bar{\tau}_f \cdot \vec{n} \\ (\bar{\tau} \cdot \vec{n})_f \cdot \vec{n} + (\rho \alpha_{eff} \nabla h)_f \cdot \vec{n} + \left\{ \left(\mu + \frac{\mu_t}{\sigma_k} \right) \nabla k \right\}_f \cdot \vec{n} \end{bmatrix}$$

Implicit Finite Volume Discretization

► Spatial Discretization

$$V_i \frac{\partial \vec{W}_i}{\partial t} + \sum_{j \in N(i)} (\vec{F}_{c,ij} - \vec{F}_{v,ij}) S_{ij} = 0$$



Implicit Finite Volume Discretization

► Time Integration

- Explicit

$$\frac{V_i}{\Delta t_i} \left(\vec{W}_i^{n+1} - \vec{W}_i^n \right) + \sum_{j \in N(i)} \left(\vec{F}_{c,ij}^n - \vec{F}_{v,ij}^n \right) S_{ij} = 0$$

- Implicit (Backward Euler)

$$\frac{V_i}{\Delta t_i} \left(\vec{W}_i^{n+1} - \vec{W}_i^n \right) + \sum_{j \in N(i)} \left(\vec{F}_{c,ij}^{n+1} - \vec{F}_{v,ij}^{n+1} \right) S_{ij} = 0$$

Implicit Finite Volume Discretization

► Linearizing Flux Vector

- Linearizing both convective and viscous fluxes using Taylor's series expansion.

$$\vec{F}_{ij}^{n+1} \approx \vec{F}_{ij}^n + \left(\frac{\partial \vec{F}}{\partial \vec{W}} \right)_{ij} \Delta \vec{W}_{ij}^n$$

Result in

$$\frac{V_i}{\Delta t_i} \Delta \vec{W}_i^n + \sum_{j \in N(i)} (A_{c,ij} - A_{v,ij}) \Delta \vec{W}_{ij}^n S_{ij} = -Res_i^n$$

where

$$\Delta \vec{W}_i^n = \vec{W}_i^{n+1} - \vec{W}_i^n$$

$$A_c = \frac{\partial \vec{F}_c}{\partial \vec{W}} : \text{Convective Flux Jacobian}$$

$$A_v = \frac{\partial \vec{F}_v}{\partial \vec{W}} : \text{Viscous Flux Jacobian}$$

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Lower-Upper Symmetric Gauss-Seidel(LU-SGS) Algorithm

► Evaluation of Flux Jacobian

- Steger-Warming's Flux Vector Splitting for Convective Flux
- Thin Shear Layer Approximation(TSL) for Viscous Flux

$$\frac{V_i}{\Delta t_i} \Delta \vec{W}_i^n + \sum_{j \in N(i)} \left(A_{c,i}^+ + A_{v,i}^* \right) \Delta \vec{W}_i^n S_{ij} + \sum_{j \in N(i)} \left(A_{c,j}^- - A_{v,j}^* \right) \Delta \vec{W}_j^n S_{ij} = -Res_i^n$$

Lower-Upper Symmetric Gauss-Seidel(LU-SGS) Algorithm

- ▶ Split off-diagonal term into lower(owner) and upper(neighbor) part

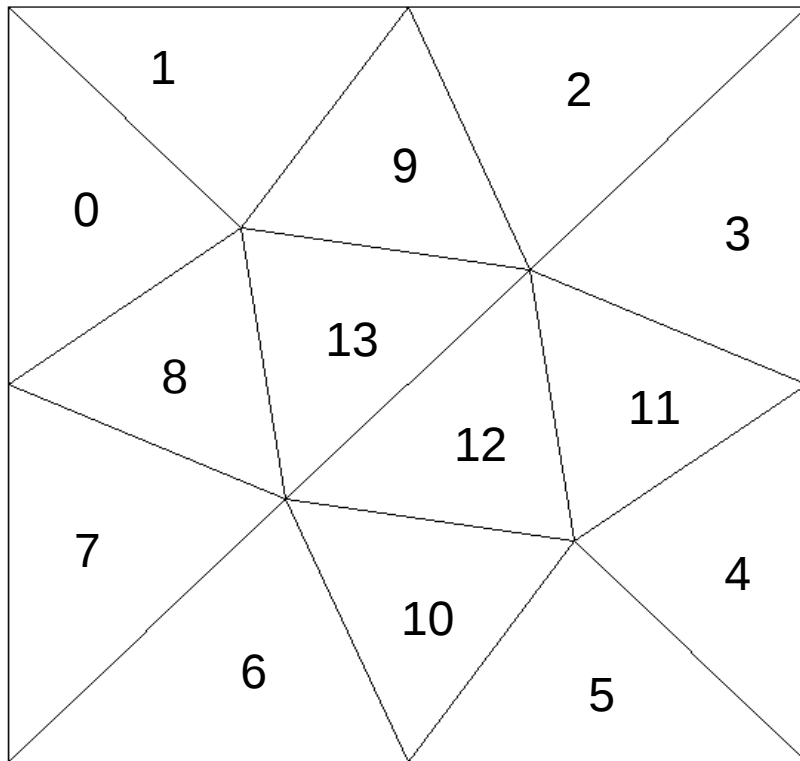
$$\begin{aligned} \frac{V_i}{\Delta t_i} \Delta \vec{W}_i^n + \sum_{j \in N(i)} \left(A_{c,i}^+ + A_{v,i}^* \right) \Delta \vec{W}_i^n S_{ij} \\ + \sum_{j \in L(i)} \left(A_{c,j}^- - A_{v,j}^* \right) \Delta \vec{W}_j^n S_{ij} \\ + \sum_{j \in U(i)} \left(A_{c,j}^- - A_{v,j}^* \right) \Delta \vec{W}_j^n S_{ij} = -Res_i^n \end{aligned}$$

- Block matrix system

$$(D+L+U) \Delta W^n = -R^n$$

Lower-Upper Symmetric Gauss-Seidel(LU-SGS) Algorithm

► Block Matrix Example



$$\begin{array}{cccccccccccccccc}
 D_0 & U_1 & 0 & 0 & 0 & 0 & 0 & 0 & U_8 & 0 & 0 & 0 & 0 & 0 \\
 L_0 & D_1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & U_9 & 0 & 0 & 0 & 0 \\
 0 & 0 & D_2 & U_3 & 0 & 0 & 0 & 0 & 0 & U_9 & 0 & 0 & 0 & 0 \\
 0 & 0 & L_2 & D_3 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & U_{11} & 0 & 0 \\
 0 & 0 & 0 & 0 & D_4 & U_5 & 0 & 0 & 0 & 0 & 0 & U_{11} & 0 & 0 \\
 0 & 0 & 0 & 0 & L_4 & D_5 & 0 & 0 & 0 & 0 & U_{10} & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & D_6 & U_7 & 0 & 0 & U_{10} & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & L_6 & D_7 & U_8 & 0 & 0 & 0 & 0 & 0 \\
 L_0 & 0 & 0 & 0 & 0 & 0 & 0 & L_7 & D_8 & 0 & 0 & 0 & 0 & U_{13} \\
 0 & L_1 & L_2 & 0 & 0 & 0 & 0 & 0 & 0 & D_9 & 0 & 0 & 0 & U_{13} \\
 0 & 0 & 0 & 0 & 0 & L_5 & L_6 & 0 & 0 & 0 & D_{10} & 0 & U_{12} & 0 \\
 0 & 0 & 0 & L_3 & L_4 & 0 & 0 & 0 & 0 & 0 & 0 & D_{11} & U_{12} & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & L_{10} & L_{11} & D_{12} & U_{13} \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & L_8 & L_9 & 0 & 0 & L_{12} & D_{13}
 \end{array}$$

$$D + L + U$$

Lower-Upper Symmetric Gauss-Seidel(LU-SGS) Algorithm

▶ Approximate Factorization

$$(D+L)D^{-1}(D+U)\Delta W^n = -R^n + LD^{-1}U\Delta W^n$$

– Factorization error

$$o(\Delta t^2) \quad \text{if} \quad \Delta t \ll \Delta x$$

$$o(\Delta t) \quad \text{if} \quad \Delta t \gg \Delta x$$

Lower-Upper Symmetric Gauss-Seidel(LU-SGS) Algorithm

► Invert the matrix in two steps

- Forward sweep

$$(D+L)\Delta W^* = -R^n$$

$$\begin{pmatrix}
 D_0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 L_0 & D_1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & D_2 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & L_2 & D_3 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & D_4 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & L_4 & D_5 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & D_6 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & L_6 & D_7 & 0 & 0 & 0 & 0 & 0 & 0 \\
 L_0 & 0 & 0 & 0 & 0 & 0 & 0 & L_7 & D_8 & 0 & 0 & 0 & 0 & 0 \\
 0 & L_1 & L_2 & 0 & 0 & 0 & 0 & 0 & 0 & D_9 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & L_5 & L_6 & 0 & 0 & 0 & D_{10} & 0 & 0 & 0 \\
 0 & 0 & 0 & L_3 & L_4 & 0 & 0 & 0 & 0 & 0 & 0 & D_{11} & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & L_{10} & L_{11} & D_{12} & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & L_8 & L_9 & 0 & 0 & L_{12} & D_{13}
 \end{pmatrix}
 \begin{pmatrix}
 \Delta W_0^* \\
 \Delta W_1^* \\
 \Delta W_2^* \\
 \Delta W_3^* \\
 \Delta W_4^* \\
 \Delta W_5^* \\
 \Delta W_6^* \\
 \Delta W_7^* \\
 \Delta W_8^* \\
 \Delta W_9^* \\
 \Delta W_{10}^* \\
 \Delta W_{11}^* \\
 \Delta W_{12}^* \\
 \Delta W_{13}^*
 \end{pmatrix}
 = -[R^n]$$

Lower-Upper Symmetric Gauss-Seidel(LU-SGS) Algorithm

– Backward sweep

$$(D+U) \Delta W^n = D \Delta W^*$$

$$\begin{pmatrix}
 D_0 & U_1 & 0 & 0 & 0 & 0 & 0 & 0 & U_8 & 0 & 0 & 0 & 0 & 0 \\
 0 & D_1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & U_9 & 0 & 0 & 0 & 0 \\
 0 & 0 & D_2 & U_3 & 0 & 0 & 0 & 0 & 0 & U_9 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & D_3 & 0 & 0 & 0 & 0 & 0 & 0 & U_{11} & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & D_4 & U_5 & 0 & 0 & 0 & 0 & U_{11} & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & D_5 & 0 & 0 & 0 & U_{10} & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & D_6 & U_7 & 0 & 0 & U_{10} & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & D_7 & U_8 & 0 & 0 & 0 & 0 & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & D_8 & 0 & 0 & 0 & 0 & U_{13} \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & D_9 & 0 & 0 & 0 & U_{13} \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & D_{10} & 0 & U_{12} & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & D_{11} & U_{12} & 0 \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & D_{12} & U_{13} \\
 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & D_{13}
 \end{pmatrix}
 \begin{pmatrix}
 \Delta W_0^n \\
 \Delta W_1^n \\
 \Delta W_2^n \\
 \Delta W_3^n \\
 \Delta W_4^n \\
 \Delta W_5^n \\
 \Delta W_6^n \\
 \Delta W_7^n \\
 \Delta W_8^n \\
 \Delta W_9^n \\
 \Delta W_{10}^n \\
 \Delta W_{11}^n \\
 \Delta W_{12}^n \\
 \Delta W_{13}^n
 \end{pmatrix}
 = [D \Delta W^*]$$

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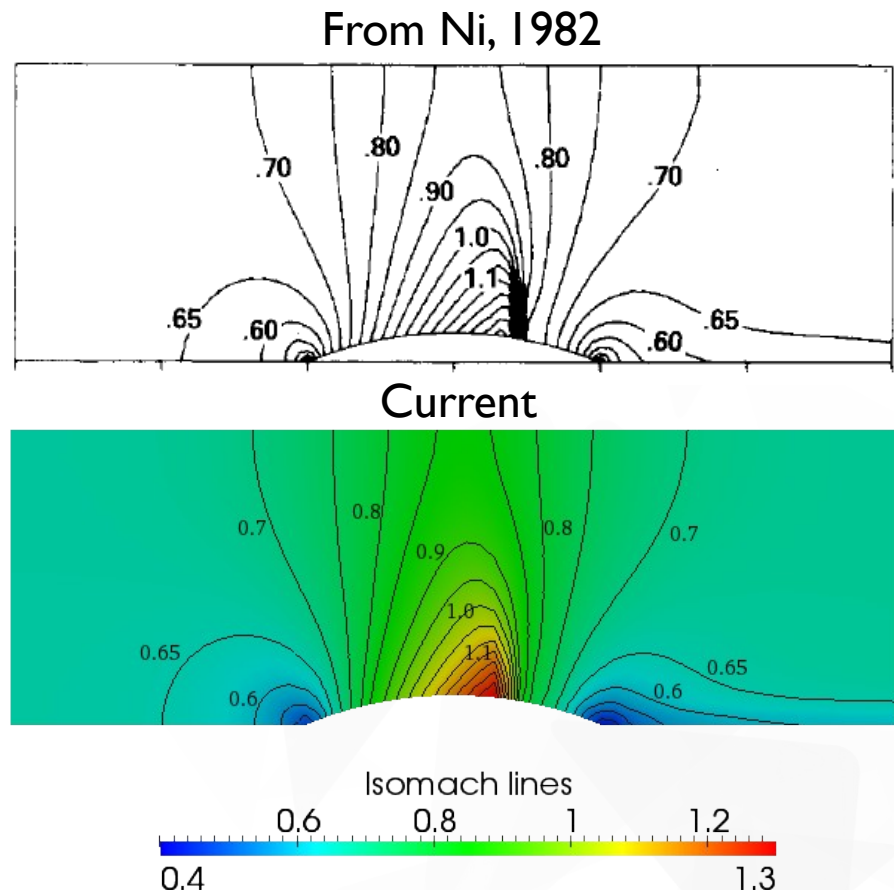
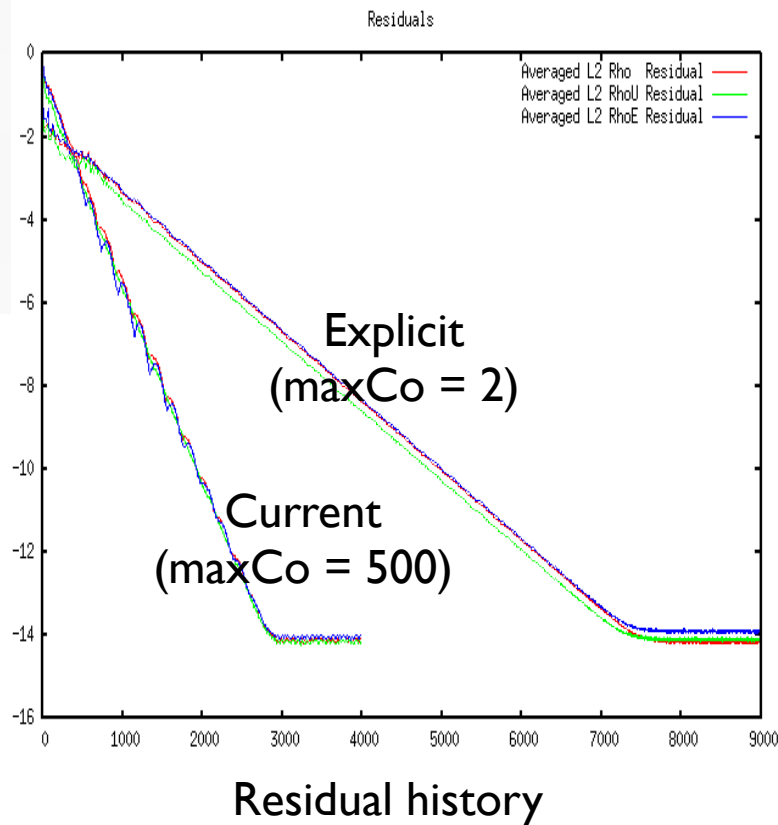
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Results – Inviscid Flow(2D)

► Transonic Flow over a Bump in Channel

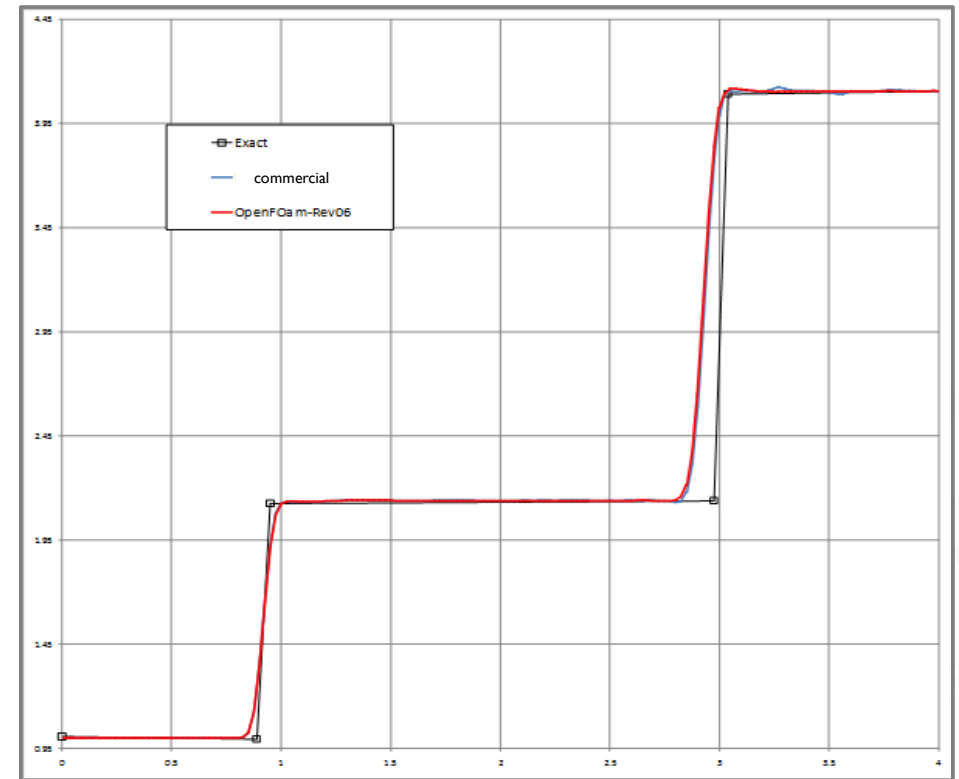
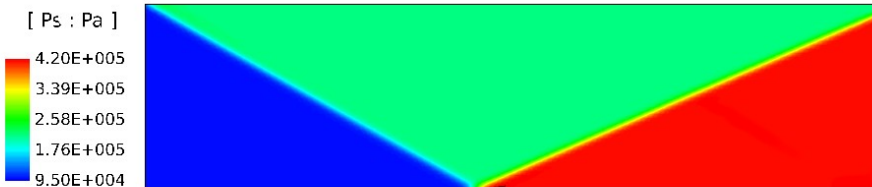
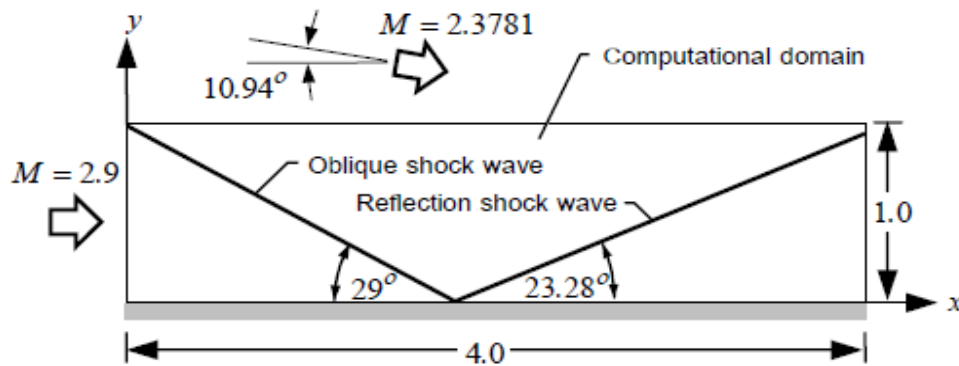
- 65 X 17 grid
- $M = 0.675$



Results – Inviscid Flow(2D)

► Oblique Shock Reflection on a Plane Wall

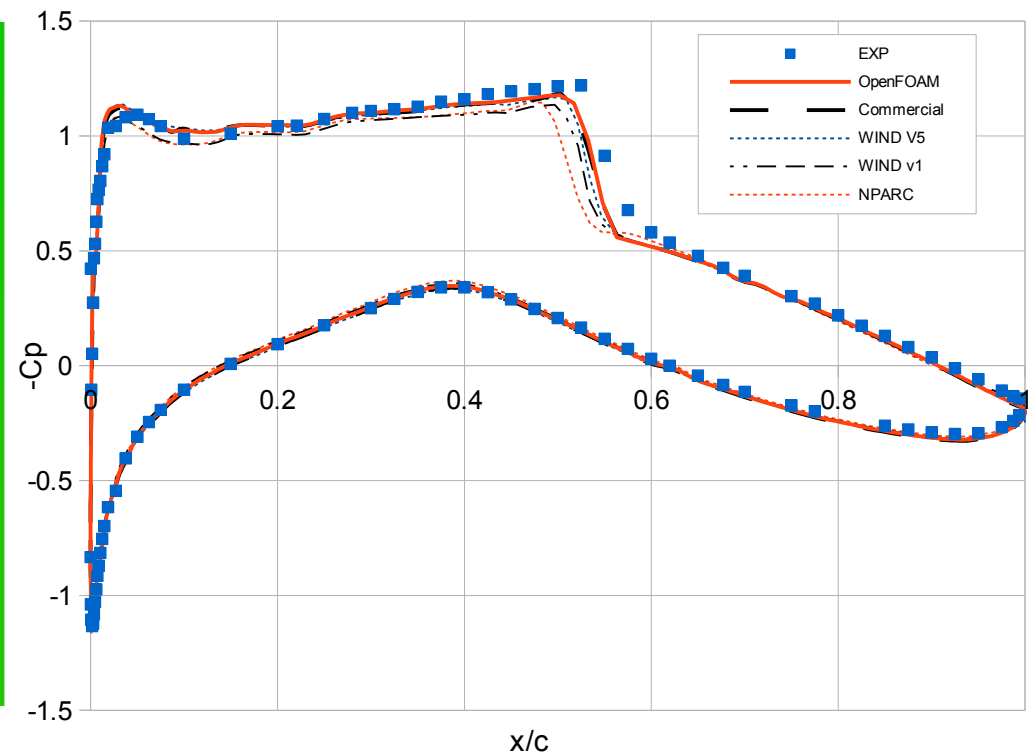
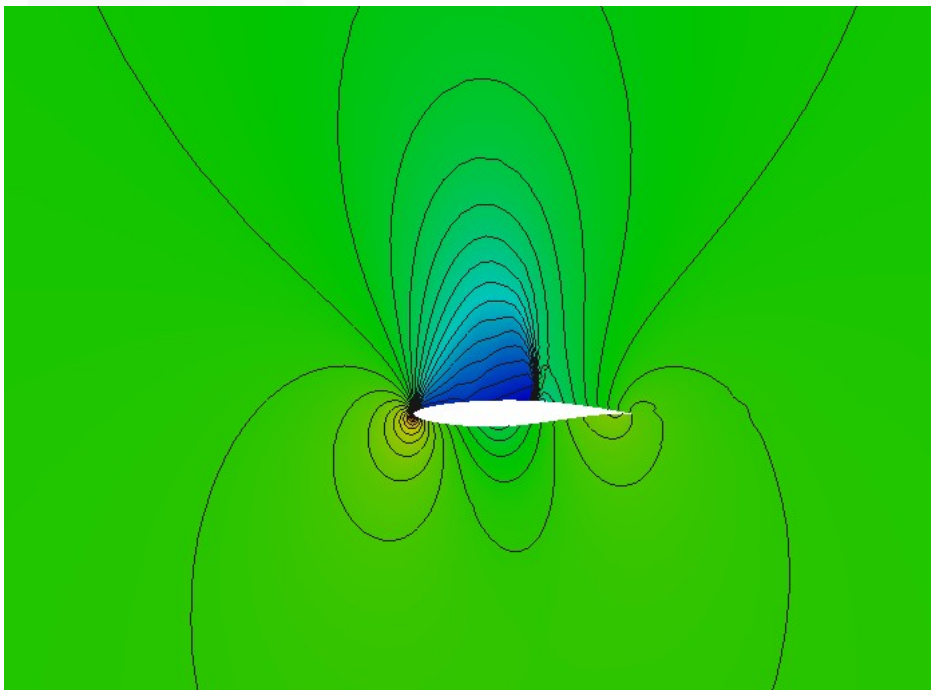
- Uniform triangular grid
- $M_\infty = 2.9$



Results – Viscous Turbulent Flow(2D)

► Transonic Flow Over RAE2822 Airfoil

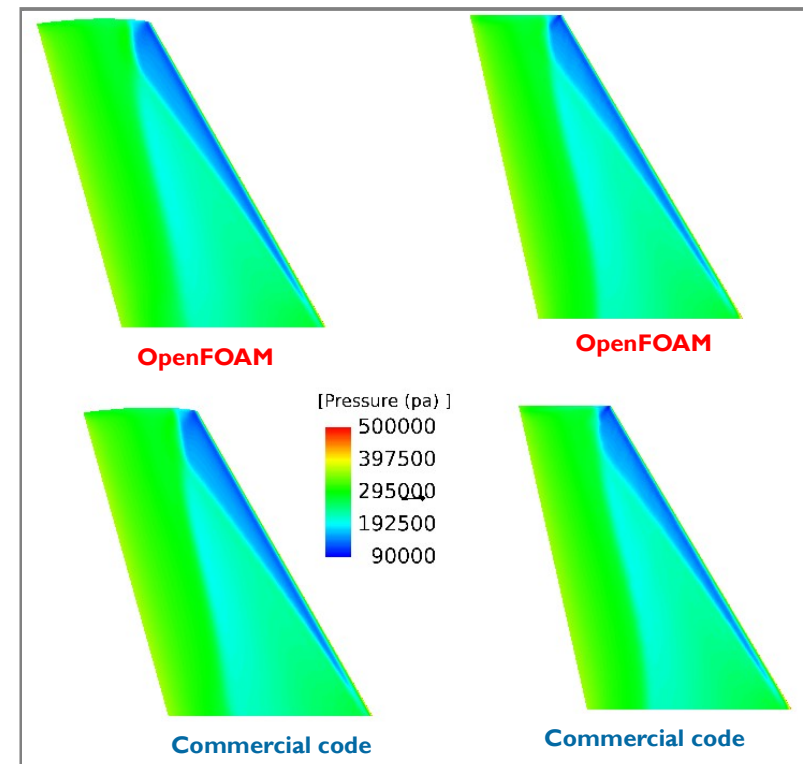
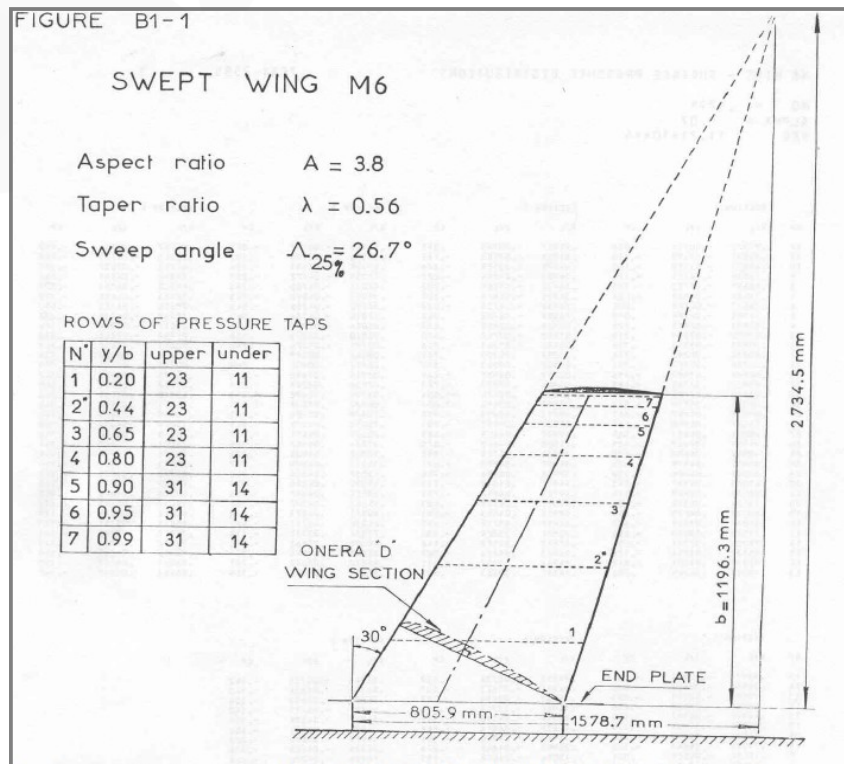
- 36,000 cells Hybrid(quad + tri) grid
- $M_\infty = 0.675$
- komegaSST turbulence model



Results – Viscous Turbulent Flow(3D)

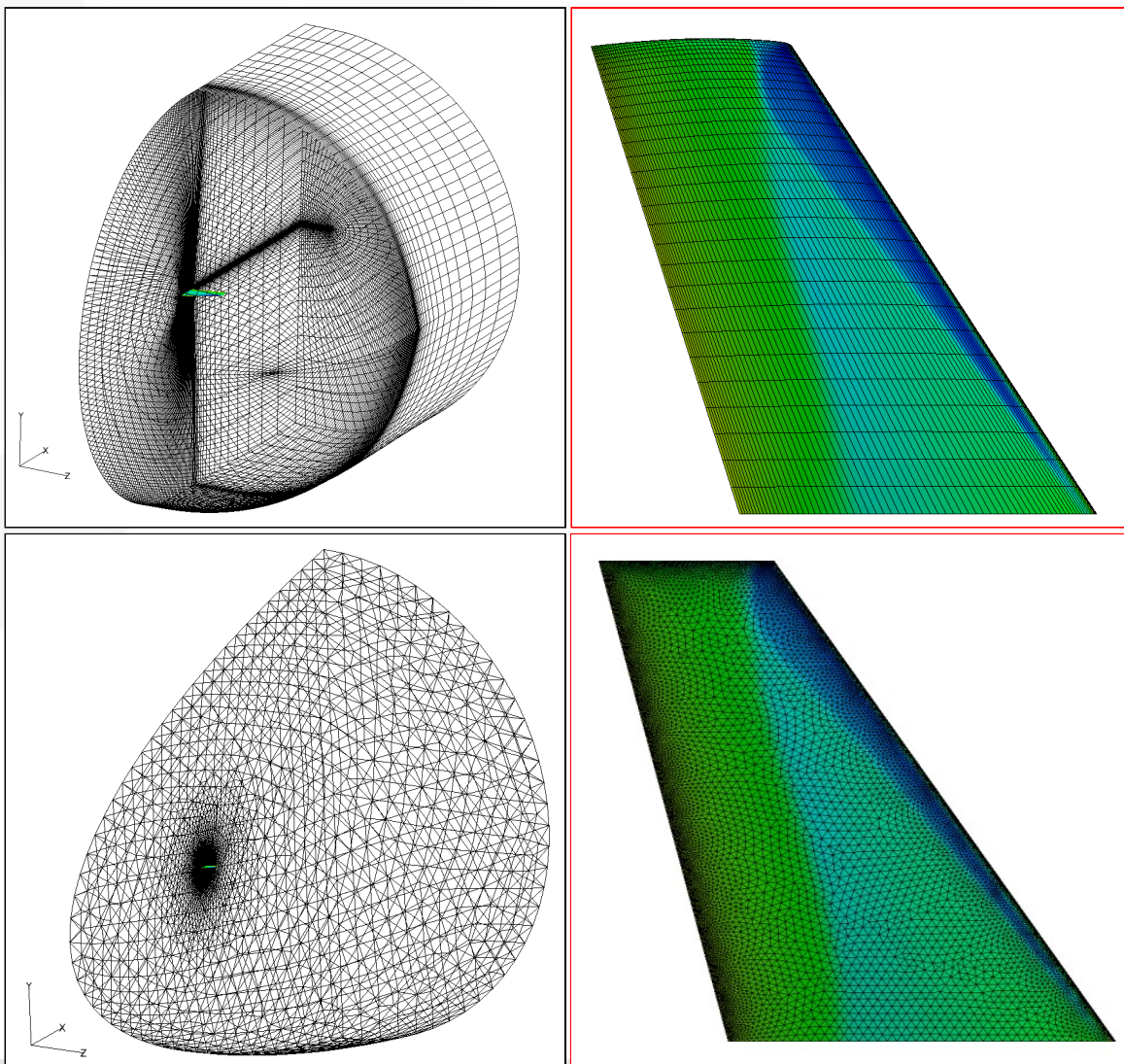
► Transonic Flow over ONERA M6 Wing

- $M_\infty = 0.8395$, $p_\infty = 315980 \text{ pa}$, $T_\infty = 255.56 \text{ K}$
- $AoA = 3.06$
- Hex(NASA) and hybrid mesh



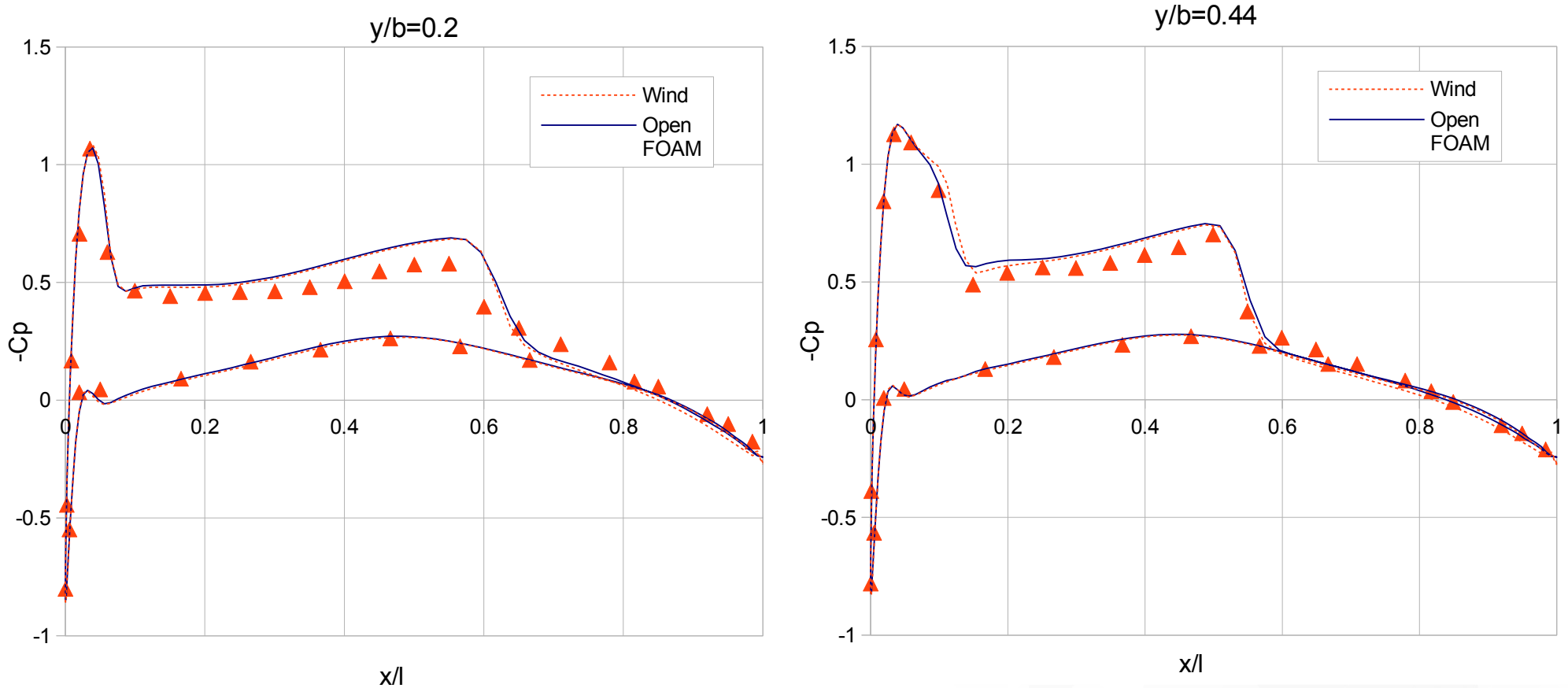
Results – Viscous Turbulent Flow(3D)

► Transonic Flow over ONERA M6 Wing



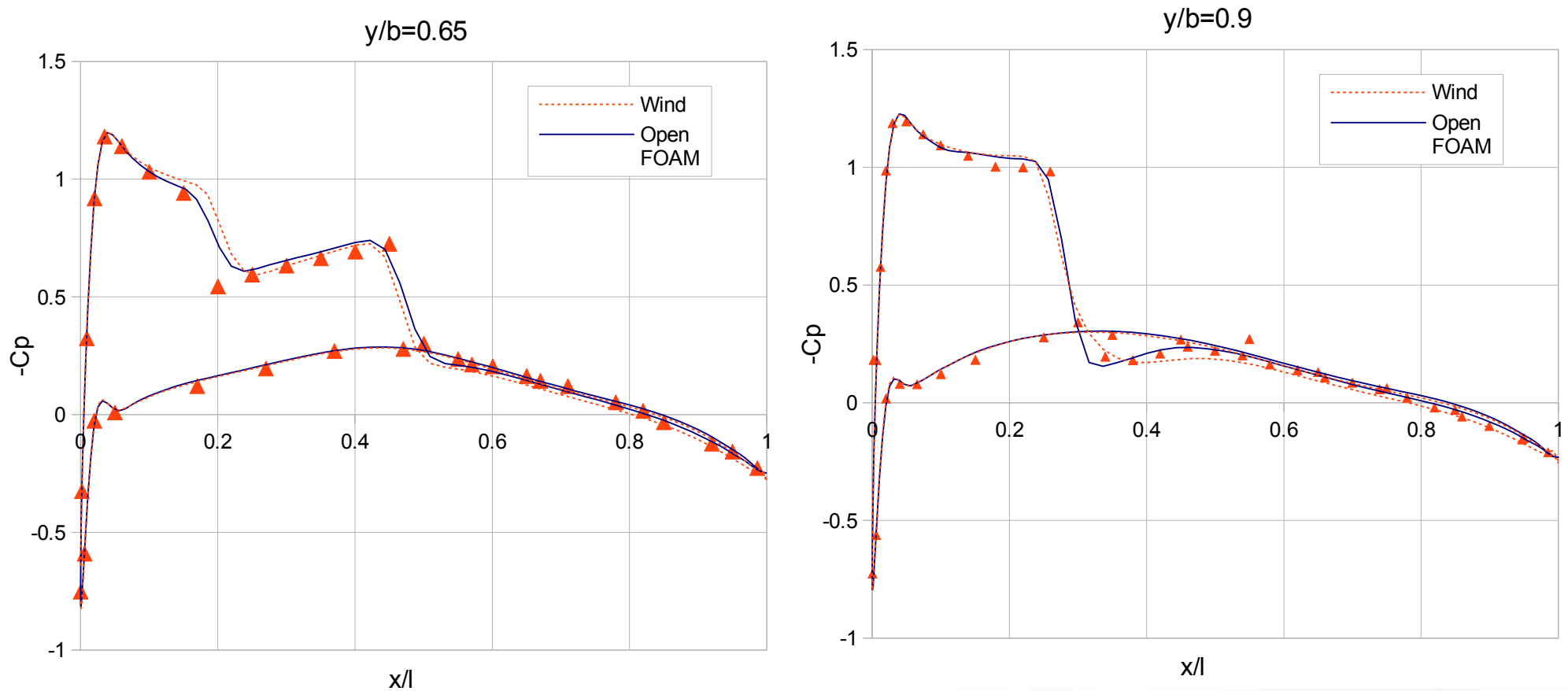
Results – Viscous Turbulent Flow(3D)

► Transonic Flow over ONERA M6 Wing



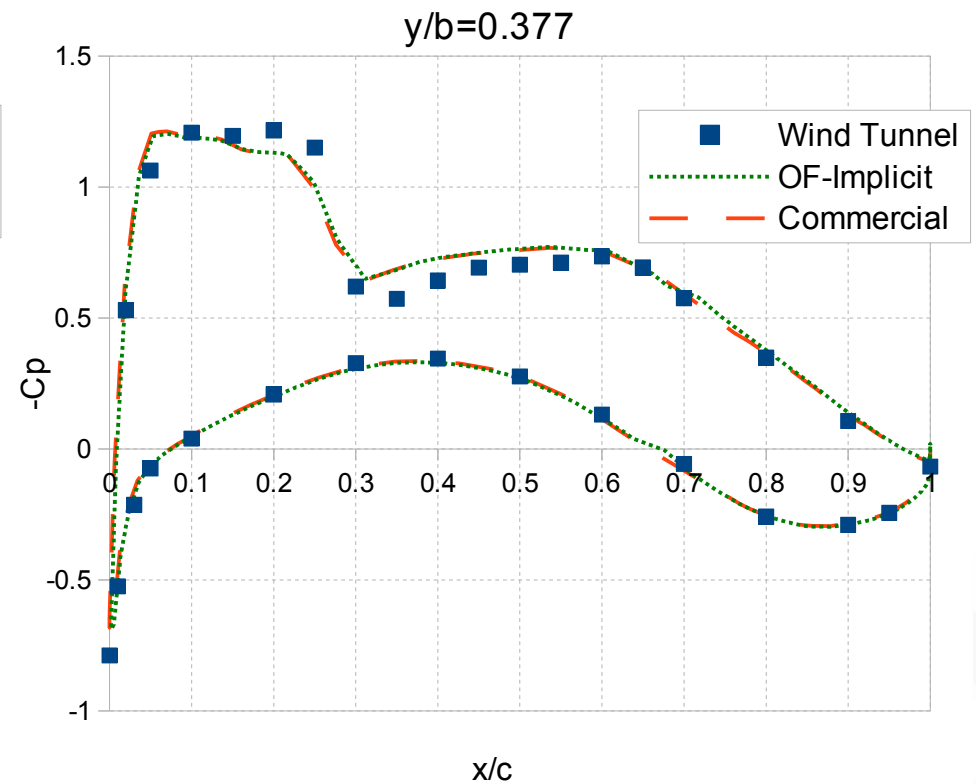
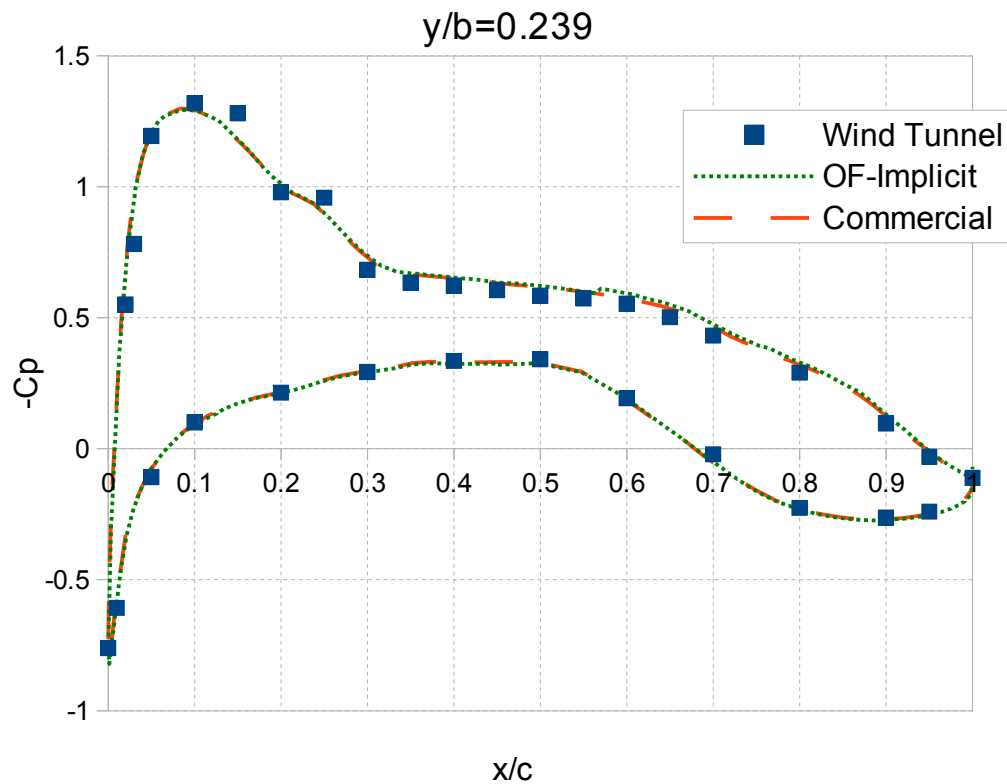
Results – Viscous Turbulent Flow(3D)

► Transonic Flow over ONERA M6 Wing



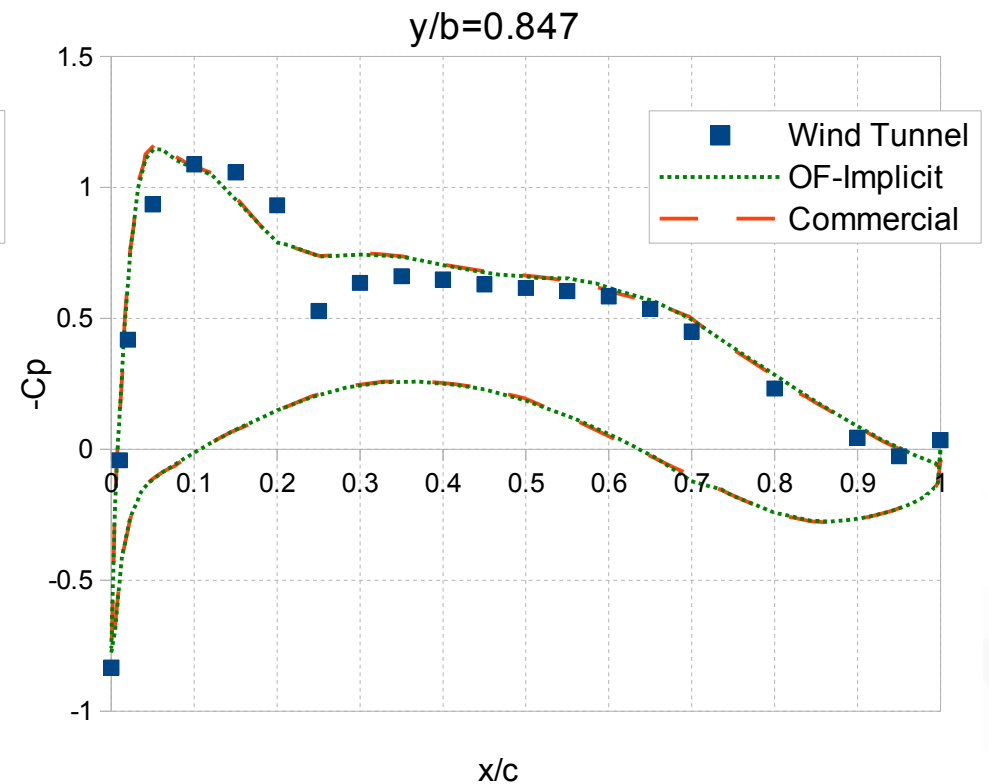
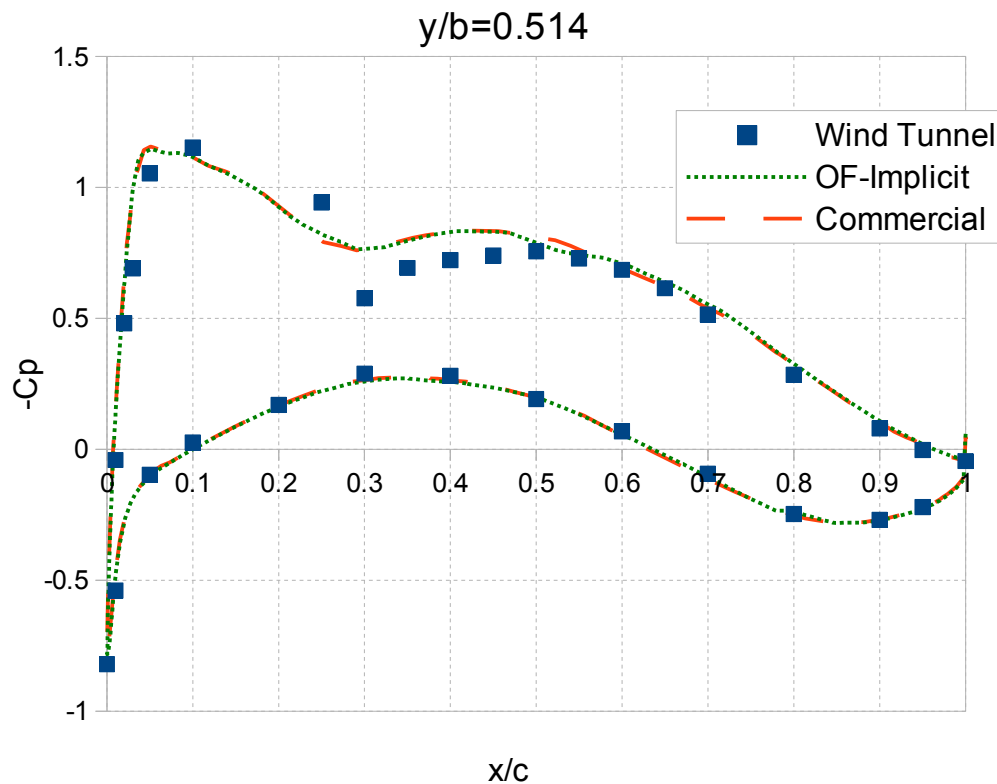
Results – Viscous Turbulent Flow(3D)

► Transonic Flow around DLR F4 Wing-Body



Results – Viscous Turbulent Flow(3D)

► Transonic Flow around DLR F4 Wing-Body



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Concluding Remarks

► Summary

- Implicit LU-SGS algorithm and new boundary condition have been implemented
- The numerical results obtained indicate that implicit algorithm leads to an increase in performance over the explicit counter part
- Solution was comparable to commercial code

► Future works

- More efficient version of LU-SGS
- Low mach number preconditioning
- Multigrid



Thank you ...

... for your attention!