Flutter analysis with an Euler-based solver in OpenFOAM

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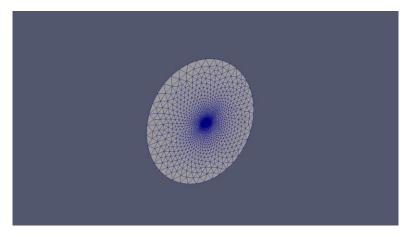


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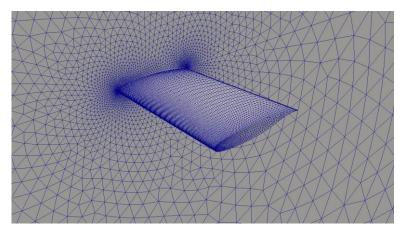
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Mesh generation



Semi-spheric domain



Wing close-up

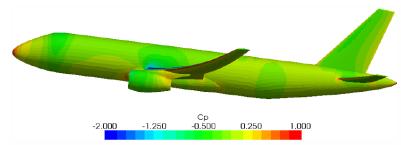
Problems in converting the provided meshes to OpenFOAM

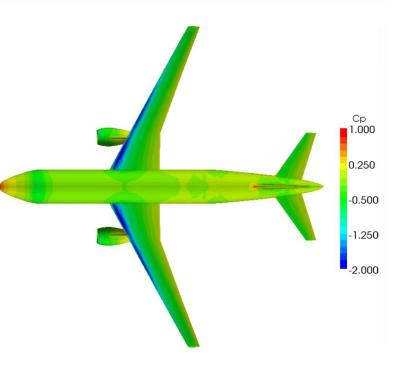
- Mesh created with Pointwise through the IGES file available on the AEPW2 project web site
- □ Spatial discretization of the domain:
 - Coarse mesh with 320k cells
 - Medium mesh with 690k cells

Aerodynamic solver: AeroFoam

In-house solver developed at the DAST supported by OpenFOAM libraries

- RANS, cell-centered, density based solver for aero-servoelastic applications
- First density-based RANS solver implemented in OpenFOAM to overcome the limits of the available pressure based solvers in transonic application





Aerodynamic solver: AeroFoam

- Euler-option is selected for the following simulations: viscosity and thermal conductivity effects are not modeled
- Convective fluxes are discretized by the Roe's approximated Riemann solver, blended by the centered approximation of Lax-Wendroff
- □ Entropy fix of Harten and Hyman and van Leer flux limiter
- Time discretization performed by an explicit multi-step Runge-Kutta scheme of the 5th order
- Combined dual-time stepping and a full-approximation storage multi-grid technique to speed up the convergence between time steps

Simulation settings

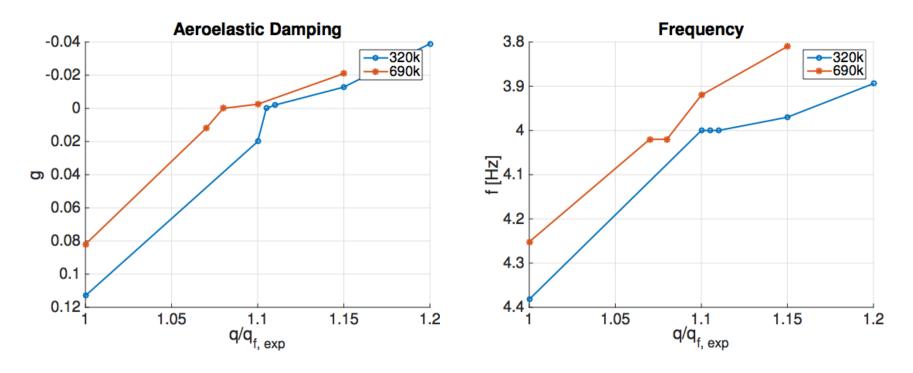
Aeroelastic interface

- Mode shapes downloaded from the AEPW2 project web site
- Because the wing is rigid, the mesh is translated and rotated rigidly during the simulation
- The coupling between structural and aerodynamic models is performed at each time step through a linear method

Aerodynamic solver parameters

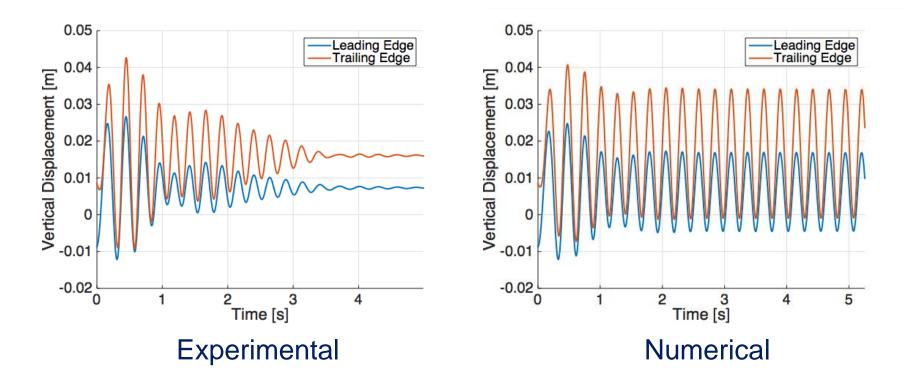
- Time step convergence analysis: 1e-3, 5e-4, 2.5e-4 s
 CFL set to 2.0
- 1000 iterations in pseudo-time are allowed between each time step
- Two multi-grid levels are used (V-cycle)

Flutter point estimation



Flutter point always overestimated vs experimental value
 Error on flutter frequency smaller than damping
 Flutter point for 320k mesh -> 1.105 experimental value
 Flutter point for 690k mesh -> 1.080 experimental value
 Flutter frequency always around 4 Hz

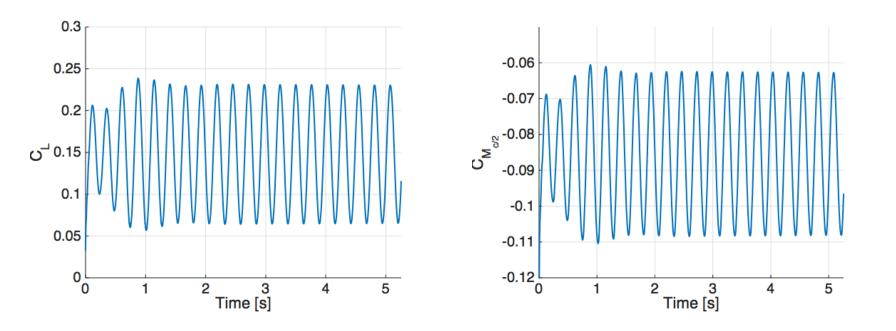
Experimental vs numerical flutter



Temporal convergence at flutter speed

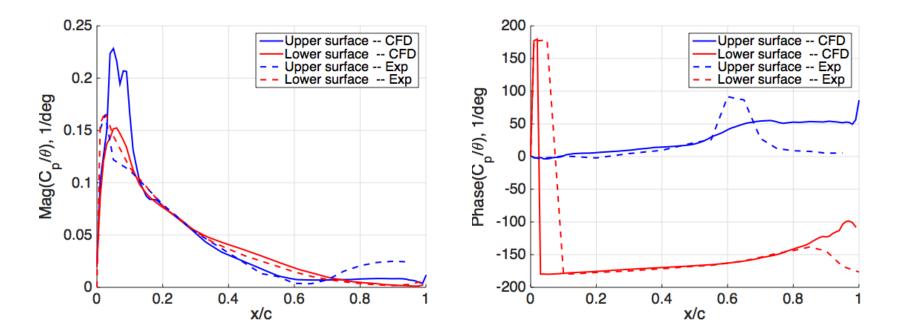
Damping variation from dt = 1e-3 to 5e-4
 No variation of damping from dt = 5e-4 to 2.5e-4

Analysis of the flutter solution



- The oscillations are not symmetric with respect to the origin
- The average angle of rotation is negative
- □ The average wing plunge is positive

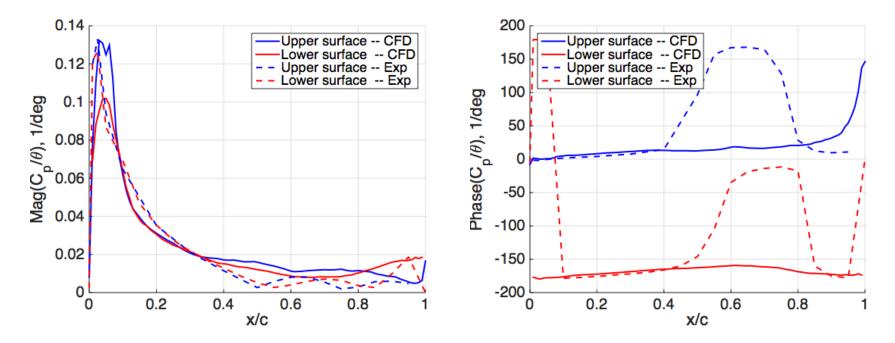
Load distribution at computational flutter



FRF @ 60% span

- The computational model presents a higher peak on the upper surface
- Phase predicted with good accuracy, missing effect at 60% chord, probably due to boundary layer transition

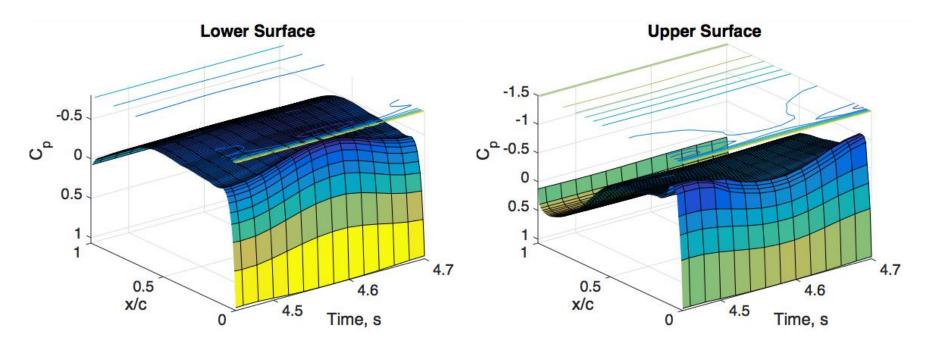
Load distribution at computational flutter



FRF @ 95% span

- The computational model presents a smaller peak on the lower surface
- Phase not well predicted, still missing effect at 60% chord, probably due to boundary layer transition

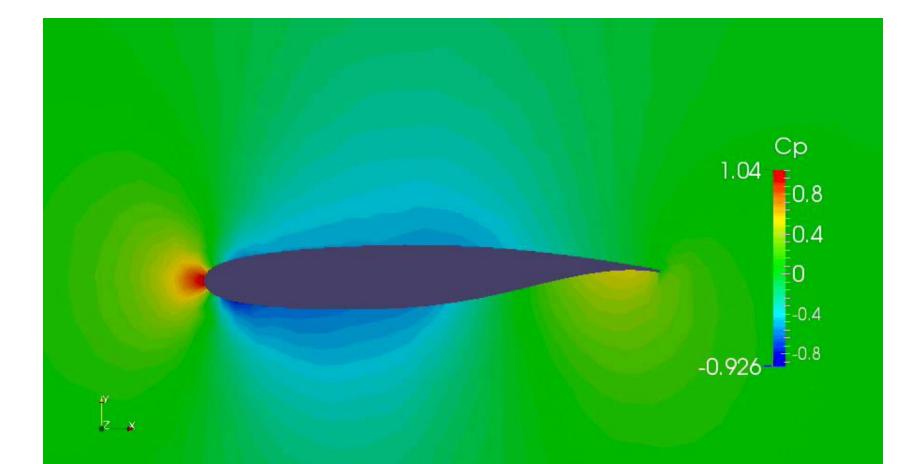
Load distribution vs time



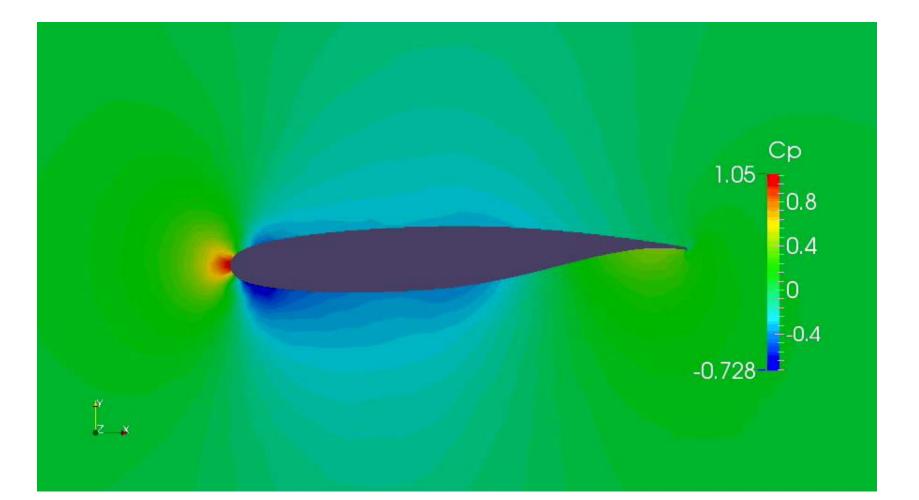
Cp @ 60% span

- A weak shock moves back and forth the chord on both surfaces
- A narrow peak is present on the leading edge of the upper surface

Pressure field @ 60% span



Pressure field @ 95% span



Concluding remarks

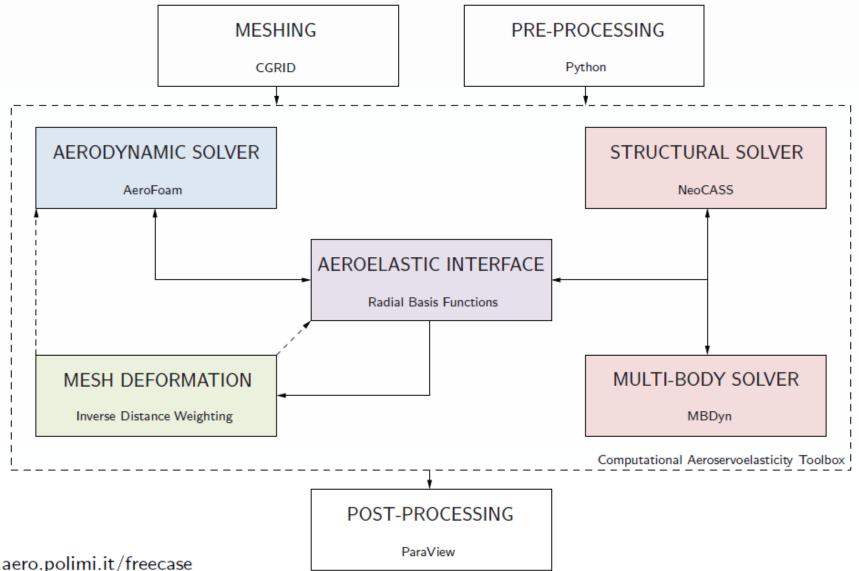
- Euler-based flutter simulations have been presented
- Good accuracy in flutter estimation (error smaller than 10%) has been found
- The flutter point is always overpredicted with respect to the experimental value
- Not so accurate in load distribution predictions, probably due to non-modeled effects
- Additional analyses with refined meshes should be carried out to confirm the convergence toward the flutter predicted by the experiments

Thank you!!! Any question?

Cases under investigation

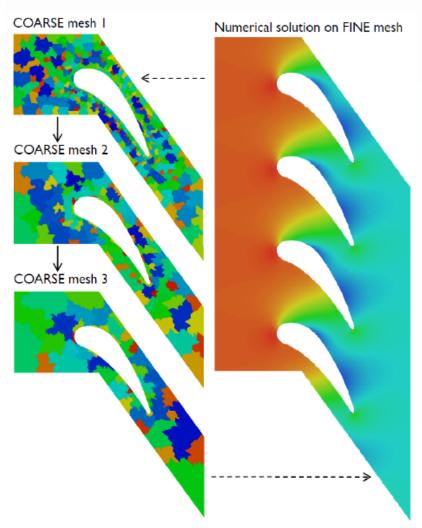
case 1) Mach = 0.7 $AOA = 3^{\circ}$ Dynamic data type = Forced oscillation, f = 10Hz, $|theta| = 1^{\circ}$ notes: attached flow, OTT exp data, R-134° case 2) Mach = 0.74 $A \circ A = 0^{\circ}$ Dynamic data type = Flutter notes = flow state unknown, PAPA exp. data, R-12 Solver: Explicit - dual time stepping density based □euler / rans (spalart allmaras / SST) grid deformation / traspiration GPU

FreeCASE toolbox



www.aero.polimi.it/freecase

AeroFoam



Multi-Grid (MG) agglomeration and Full-Approximation-Storage (FAS)

Motivation and objective:

- a) First density-based ALE RANS solver in OpenFOAM
- b) To overcome the limits of built-in pressure-based solvers in the transonic regime (e.g. sonicFoam)
- c) Benchmarking vs. EDGE and FLUENT

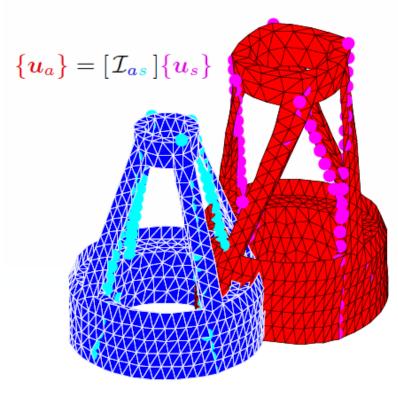
Features:

- a) Coupled formulation in conservative variables
- b) Space discretization (1st, 2nd order accuracy)
 - Roe's Approximate Riemann Solver (ARS)
 - Lax-Wendroff (LW) scheme with flux limiters
 - Directional Residual Smoothing (RS)
- c) Time discretization (1st, 2nd order accuracy)
 - Explicit multi-step Runge-Kutta (RK) scheme
 - Local Time-Stepping (LTS)
 - Double Time-Stepping (DTS)
 - Multi-Grid (MG) acceleration (FMG and FAS options)
- d) Automatic handling of parallel communication, cyclic boundaries, Generic Grid Interface (GGI)

FSI

Target: closed loop connection between structural and aerodynamic sub-systems

- a) projection (in PVW sense) of aerodynamic forces onto structural displacements
- b) translation of structural displacements into aerodynamic boundary conditions



Moving Least Squares (MLS):

- connect topologically different domains
- exact treatment of rigid motions
- accuracy, smoothness & efficiency trade-off

$$\mathsf{Minimize} \ \int_{\Gamma} \phi(\operatorname{Tr}(\boldsymbol{u}_a)|_{\Gamma} - \operatorname{Tr}(\boldsymbol{u}_s)|_{\Gamma})^2 \mathrm{d}\mathcal{S}$$

• weighting via Radial Basis Functions (RBF)

A hierarchy of mesh deformation tools

R Least-squares identification of translation vector \mathbf{s} and linear map tensor \mathbf{T}

$$\Delta \boldsymbol{x}_j = \boldsymbol{s} + \boldsymbol{\mathsf{T}} \, \boldsymbol{x}_j + \boldsymbol{\varepsilon}_j = \boldsymbol{s} + (\boldsymbol{\mathsf{R}} - \boldsymbol{\mathsf{I}}) \, \boldsymbol{x}_j + \boldsymbol{\mathsf{D}} \, \boldsymbol{x}_j + \boldsymbol{\varepsilon}_j \quad \forall \quad j \in [1, \, N_b]$$

easier to implement, rotation tensor follows: $(\mathbf{R} - \mathbf{I}) = s_{\phi} \mathbf{K}_{\times} + (1 - c_{\phi}) \mathbf{K}_{\times} \mathbf{K}_{\times}$

E Elastic contribution by means of Sparse Inverse Distance Weighting (SIDW)

$$\Delta \boldsymbol{x}_{\boldsymbol{k}} = \sum_{j=0}^{N_b} \frac{\mathsf{IDW}_{(\boldsymbol{k},j)}}{|\mathsf{IDW}_{(\boldsymbol{k},j)}|} \boldsymbol{\varepsilon}_{\boldsymbol{j}} \quad \forall \quad \boldsymbol{k} = [1, N_v] \quad \text{with} \quad \mathsf{IDW}_{(\boldsymbol{k},j)} = \frac{1}{\|\boldsymbol{x}_{\boldsymbol{k}} - \boldsymbol{x}_{\boldsymbol{j}}\|^p}$$

memory/efficiency trade-off sparse fix: $SIDW_{(k,j)} = IDW_{(k,j)}$ if $IDW_{(k,j)} > \xi$

T Residuals (if any) simulated by means of Transpiration boundary conditions