An advanced CFD postprocessing tool enabled by adjoint and RBF mesh morphing allows the fast exploration of an arbitrary number of shape parameters effect on performances

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Outline

- RBF Morph technology and software solutions for **shape optimization**
- Gradient based examples based on **adjoint** solver
- Beyond (before) optimization: advanced **post processing**
- Industrial example: exploring the drag reduction of a **Formula Indy** car
- **Conclusions**

10+ years of RBF Morph

Industries served (100+ institutions)

Automotive

Aerospace & Defence

Healthcare & Medical

Nautical & Marine

Energy

Radial Basis Functions mesh Morphing

- We offer **Radial Basis Functions** (RBF) to drive mesh morphing (smoothing) from a list of source points and their displacements
	- **Surface shape changes**
	- Volume mesh smoothing
- RBF are recognized to be one of the **best mathematical tool** for mesh morphing

$$
s_x(x) = \sum_{i=1}^{N} \gamma_i^x \varphi(||x - x_{s_i}||) + \beta_1^x + \beta_2^x x + \beta_3^x y + \beta_4^x z
$$

$$
s_y(x) = \sum_{i=1}^{N} \gamma_i^y \varphi(||x - x_{s_i}||) + \beta_1^y + \beta_2^y x + \beta_3^y y + \beta_4^y z
$$

$$
s_z(x) = \sum_{i=1}^{N} \gamma_i^z \varphi(||x - x_{s_i}||) + \beta_1^z + \beta_2^z x + \beta_3^z y + \beta_4^z z
$$

Biancolini, M. E. (2017). *Fast radial basis functions for engineering applications*. Springer International Publishing.

Parametric CAE models

CAE models supported includes flow analysis (CFD) and structural analysis (FEM)

RBF Morph makes the CAE model parametric with respect to the shape.

Works for any size of the mesh.

Shape parameters can be steered with the optimizer of choice.

- It's **easy and fast**: shape parameters are defined in the CAE GUI. No need to iterate the CAD
- **The turnaround time of the optimization is usually** reduced by a factor five (weeks become days)

We offer Ansys integrated solutions…

ACT Extension (FEM)

- Released in **2014**
- Fully embedded in ANSYS **Mechanic[al](https://catalog.ansys.com/product/5b3bc6857a2f9a5c90d32ea9/rbf-morph-act-exte)** (parametric)
- **Benefits of underlying geometry** (or aux geo with dead meshes) ■ …**WB** Meshing

Fast RBF mesh morphing technology that makes the mesh shape parametric with a few clicks. Basic and hierarchical shape modifications defined in the tree. Automatic shape optimisation now included.

Meshing

Fluent Module (CFD)

- Released in **2009**
- **Fully integrated** within Fluent (GUI, TUI & solving stage), Workbench and **Adjoint Solver**
- Multi physics features (**FSI**)

…and a Stand Alone software

■ Released in 2011

- **Read in STL and CGNS file formats**
- Solver independent process that supports **many mesh formats**
	- **Scriptable** via tcl

RBF Morph Fluent Module

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Fluent module

- Add on fully integrated within Fluent (GUI, TUI & solving stage), **Workbench** and **Adjoint Solver**
- Mesh-independent RBF fit used for **surface** mesh morphing and **volume** mesh smoothing
- **Parallel** calculation allows to morph large size models (many millions of cells) in a short time
- Management of **every kind of mesh** element type (tetrahedral, hexahedral, polyhedral, etc.)
- Support of the **CAD** re-design of the morphed surfaces
- Multi fit makes the Fluent case **truly parametric** (only 1 mesh is stored)
- Precision: exact nodal movement and exact feature preservation (**RBF are better than FFD**)

Gradient-based optimization (adjoint)

Optimization with parameters

 The adjoint formulation provides the **gradient** of an aerodynamic objective function **with respect to surface displacements**

$$
\frac{\delta F}{\delta \vec{b}} = \frac{\delta F}{\delta x_{\kappa}} \frac{\delta x_{\kappa}}{\delta \vec{b}}
$$

RBF Morph provides the deformation velocity (adjoint preview) δb δx_{κ} δb
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Engine Air box

-
- **32 shape parameters** are used to control the geometry of the plenum and of the three runners
- Obtained shape allows to get a **15.3%** reduction of pressure drop and uniform distribution

Engine Air box

- **V12 700Hp@8250RPM**
- Obtained shape allows to get a **5.9%** reduction of pressure drop

Beyond (before) optimization: advanced adjoint based post-processing

- **High fidelity CAE solver adopted in this** study is **Ansys Fluent** (CFD+adjoint)
- Advanced mesh morphing is provided by combining the CFD solver with the **RBF Morph Fluent module**
- A new **interactive custom feature** defined to quickly explore new shapes without any additional solver calculations **(rbf-adjoint-interactive)**

Beyond (before) optimization: advanced adjoint based post-processing

- 1. Inspect flow solution and adjoint sensitivity
- 2. Decide the regions to be modified
- 3. Create desired shape modifications (design parameters, FEA deflections, sculpted shapes)
- 4. Explore how shape modifications combines and get a (gradient based) estimation of the performance

- Half car model comprised of about **80 millions cells**
- **Drag sensitivity computed** with adjoint solver
- Regions of interest:
	- **Windshield**
	- Rear wheel deflector

- Shape modifications are defined to get a variety of feasible new shapes
	- **Windshield** is controlled by 9 different actions
	- **Rear wheel deflector** is controlled by 3 different actions

- Shape modifications are defined by **filtering adjoint** sensitivity field that allows to get a new optimal shape for
	- **Nindshield**
	- Rear wheel deflector

Drag analysis of a challer analymical equation

$\begin{picture}(130,10) \put(0,0){\line(1,0){10}} \put(15,0){\line(1,0){10}} \put(15,0){\line($ \circ $\overline{\bullet}$ $\hfill \Box$ E O \equiv \circ **L**

Console

Morph preview on 2 ID(s), with 0 total points... Morph preview on 2 ID(s), with θ total points... Objective function variation: θ Corresponding to a drag improvement equal to 0 drag counts

solve/ surface/ views/

all

 \bullet 0 selected

The challenge

- **Well, now we have a quantitative estimation** of the effect of shape parameters
- **Automatic methods are based** on the **gradient** and can converge toward an optimum
- Can the **engineer** get a better shape in a single shot?

…to answer…

- *Sensitivities are valid only around the baseline*
- *Too much variation could be risky*
- *A small variation is safer… but predicts a small gain!*
- *(challenge posed on Friday to get the answer after the week end)*

Predicted result

- The new shape was defined combining the 3 parameters (2 on the windshield, 1 on the rear wheel deflector) that show most promising results
- A **0.5% reduction** of the drag force has been predicted
- The new model has been generated and submitted for a full CFD run

Console

608409 polygons displayed.

Objective function variation: 0 Corresponding to a drag improvement equal to 0 drag points

> (rbf-adjoint-interactive '((sidepod_5 -0.3)(cupolino_cyl_1 0.2)(cupolino_cyl_2 -0.2)))

Predicted result

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Console

Aujuant Everlestion - 77.3173

Morphing-Adjoint evaluation done in 0 sec.

Sequential Morph Adjoint done in 0 sec. /display/surface-grid (48 51 55 56 54 52 53 59 57 58 60 47 63 31 30 41 40 43 21 22 23 44 17 45 16 18 15 20 36 34 35 19 39 37 38 > MaxDisp = 0.015002 MaxSkew = 6.329179e-01 608409 polygons displayed.

Objective function variation: -7.9179111 Corresponding to a drag improvement equal to -0.25854403 drag points

Drag analysis of a challer analymical example 2 and 2

Obtained result

- A **1% reduction** of the drag force has been obtained!
- A **0.49 drag points** reduction $(0.5392 \rightarrow 0.5343)$ achieved
- **The breakdown of drag distribution** has been inspecte

Drag analysis of a challer analymical example 2 and Sequence 2 and Sequ

Original shape (Cd= 0.5392) Optimized shape (Cd= 0.5343)

Drag analysis of a challer analyzis Analyzis Card Example 2 are:

Original shape (Cd= 0.5392)

Drag analysis of a challer analyzis Analyzis Card Example 2 are:

Optimized shape (Cd= 0.5343)

Drag analysis of a challer analymical lindy car

Original shape (Cd= 0.5392)

Optimized shape (Cd= 0.5343)

Drag analysis of a challer analymical lindy car

Original shape (Cd= 0.5392)

Optimized shape (Cd= 0.5343)

Conclusions

- **There is a need for advanced tools to get as more information as** possible from high fidelity CFD
- When **shape sensitivities** are available (adjoint solution) we can compute derivatives of performance vs. parameters
- In this study we presented (rbf-adjoint-interactive) a new tool based on Ansys Fluent, and RBF Morph that allows to predict the effect of shape on performance **without** the need of a new **CFD computation**
- The proposed method was applied to reduce the aerodynamic drag of a **Dallara Indy Car** acting on the windshield and on the rear wheel deflector gaining a **0.49 drag points**

Thank you!

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