# AUTOMATIC CFD ANALYSIS METHOD FOR SHAPE OPTIMIZATION

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**ENGINEERING SCHOOL** 

MECHANICAL ENGINEERING DEPARTMENT

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# Graduation project for the degree of Mechanical Engineer

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To my family, Luis Alberto, Luz Aida and Ana Catalina; all their faith and support has brought me here. I hope tha I can repay you someday for the sacrifices you made for me. All that I am or will become came from you and to you it will return.

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### **ABSTRACT**

This project presents an Automatic Computational Fluid Dynamics (CFD) analysis method for shape optimization of an aerodynamic profile. It begins with an overview of basic concepts on shape optimization, geometry parameterization and objective functions. It continues with an introduction to the current status of CFD simulation software and types of solvers. Then expands to optimization based on CFD analysis. Following, a CFD-based method to optimize aerodynamic profiles under certain restrictions and scenarios is proposed. Finally, the code implemented to automatically modify a profile bound by a set of control points based on CFD analysis is described.

The project was developed entirely at the EAFIT University's Applied Mechanics Laboratory in Medellin, Colombia and is part of a collaboration effort in companionship with the University of Aberta in Canada and Los Andes University in Bogota, Colombia.

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### 0 STATE OF THE ART

The following chapter presents some relevant information to this study, from which knowledge necessary for the development of the project can be obtained.

### 0.1 SHAPE OPTIMIZATION

0.1.1 What is Shape Optimization. Optimization in a broader sense is maximizing or minimizing some function relative to some set, often representing a range of choices available in a certain situation. The function allows comparison of the different choices for determining which might fit better to the selection criteria (ROCK-AFELLAR, 2007).

The maxima or minima of the function represent an important feature of the structure (s). For example in a wing airfoil it can represent the minimum drag Coefficient  $C_d$  for a given configuration. Then, if:

$$f(s) = \min C_d \tag{1}$$

The interest is to find the minimum of f(s), that is to find a  $s^*$  such that:

$$f(s^*) = min\{f(s), \forall s \in \mathbf{V}\}$$
(2)

where **V** is the space of all the admissible geometries.

Shape optimization is within the field of optimal control theory. It is the process of reaching an optimal shape through iteration over some design parameters. The optimization process couples a geometry definition and analysis code in an iterative process to produce optimum designs subject to various constraints (SONG y KEANE, 2004). These constraints allow us to find a bounded set that defines the optimal shape.

Shape optimization results rely greatly on the accuracy and relevancy of the chosen design parameters. These parameters ensure that an overall optimum is achieved. Systems using multiple design variables are more demanding in terms of computation time. Here low fidelity modeling might offer an alternative to explore design in a more conceptual fashion. Coarser approximations based on low fidelity modeling offer valuable information in situations where little initial knowledge is available (DYE et al., 2007).



Figure 1: CFD-Based optimization result of a turbine runner.

(WU et al., 2007)

0.1.2 Shape parameterization. Geometry parameterisation methods have attracted renewed interests in recent years, especially in the context of multidisciplinary design

optimisation (MDO). Samareh (SAMAREH, 2001) identified three categories of parameterization methods in the context of MDO. These include the discrete approach, CAD-based approaches, and free-form deformation methods.

To model any system, it is essential to incorporate physics and geometry-based parameterizations that are representative of the underlying "real-world" physics of the system's aerodynamics, thermodynamics, and mechanical behavior (DYE et al., 2007). Shape parameterization reduce significantly design cycle times. Having a parametric and programmatic approach makes the optimization process more effcient in terms of calculation times, parameter input tasks and computational manipulation of the data. The definition of a geometry using simple parameters that yields a detailed definition of a surface model is the ideal for shape parameterization.

It is a common practice to define geometries using curves such as bezier curves (see figure 2), b-spline curves and non-uniform rational b-splines (NURBS). Bezier curves allow an easy representation of airfoil upper and lower portions (DÉSIDÉRI y JANKA, 2004). Bezier curve parameterization can be defined as:

$$x(t) = \sum_{k=0}^{n} B_n^k(t) x_k \; ; \; y(t) = \sum_{k=0}^{n} B_n^k(t) y_k$$
 (3)

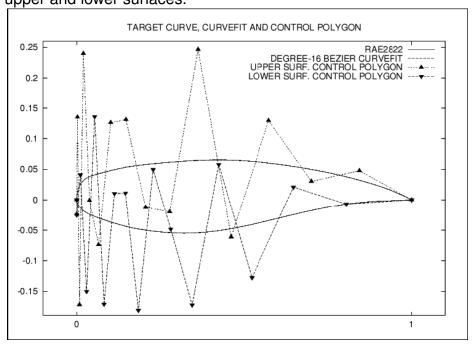
in which the parameter t varies from 0 to 1, n is the degree of the parameterization, and

$$B_n^k(t) = C_n^k t^k (1 - t)^{n - k} \tag{4}$$

is a bernstein polynomial,  $C_k^n=\frac{n!}{k!(n-k)!}$ , and  $P_k=(x_k\;y_K)\;(k=0,1,..,n)$  is a generic control point.

Another concern while selecting the type of parameterization is the grid sensitivity. The design variables will be evaluated on a CFD environment. Depending on the

Figure 2: RAE2822 airfoil, degree-16 Bézier curvefit and control polygons associated with the upper and lower surfaces.



(DÉSIDÉRI y JANKA, 2004)

complexity of the geometry the grid (mesh) quality will reflect the accuracy of the values obtained when the objective function is evaluated. Surface forces on an airfoil will be higly sensible to this aspect. For this matter the use of NURBS is recommended. The type of parameterization should be selected from an analysis of the sensitivity of the grid. The values of the state variables (governing equations) as well as the design parameters should be a key aspect to select parameterization, otherwise gradient errors could be induced (SADREHAGHIGHI et al., 1995).

Using CAD-Based parameterizations has become a common practice. The mathematics of the geometric aspects are handled by the CAD software. The manipulation of NURBS or B-Splines to describe profiles, surface or bodies are trusted to the geometric kernels of these softwares. Nowdays programmable interfaces are available

for most of this softwares to enable code integration (DYE et al., 2007).

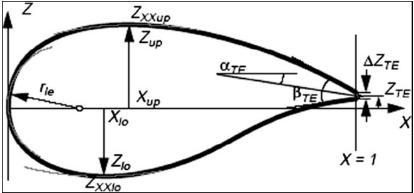
One of the key issues in deciding the parameterization method is to balance the requirements of robustness and flexibility, and these decisions are also strongly dependent on the goal of the design activity. Although some parameterization methods may well be able to generate radical new shapes, this might not be suitable for designs where the aim is to meet a specific pressure distribution or drag coefficient. Increased degrees of freedom in the control parameters -or elevated number of themwill lead to a poor efficiency caused by the large search space that arises in the optimization process (SONG y KEANE, 2004).

PARSEC is one of the most common and effective methods for airfoil representation in the design optimization field. Fig. 3 illustrates the eleven basic parameters of PARSEC method, which are the leading edge radius  $(r_{LE})$ , upper and lower crest location  $(X_{UP}, Z_{UP}, X_{LO}, Z_{LO})$  and curvature  $(Z_{xxUP}, Z_{xxLO})$ , trailing edge coordinate  $(Z_{TE})$  and direction  $(\alpha_{TE})$ , trailing edge wedge angle  $(\beta_{TE})$  and thickness  $(\Delta Z_{TE})$ . A linear combination of shape functions is used to present the airfoil shape in this method:

$$Z_k = \sum_{n=1}^{6} a_{n,k} X_k^{\frac{n-1}{2}} \tag{5}$$

Eleven design parameters are required for the PARSEC representation to completely define an airfoil shape. The coefficients  $a_n$  are determined from defined geometric parameters. The airfoil is divided into upper and lower surfaces and the coefficients  $a_n$  are determined using the information of the points in each section. The subscript k changes from 1 to 2 in order to consider the length on the upper and lower surfaces, respectively. Using the parameters mentioned above, one can effectively control the maximum curvature of the upper and lower surfaces and their location

Figure 3: PARSEC Method Parameters.



(SHAHROKHI y JAHANGIRIAN, 2007)

that are very useful in reducing the shock wave strength or delaying its occurrence. However, at the trailing edge of the airfoil, PARSEC fits a smooth curve between the maximum thickness point and the trailing edge which in turn disables the necessary changes in the curvature close to the trailing edge. Therefore, in spite of its benefits on controlling the important parameters on the upper and lower surfaces, PARSEC does not provide enough control over the trailing edge shape where important flow phenomena can occur.

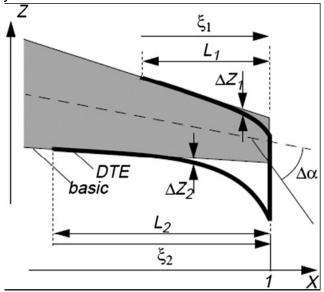
A technique for overcoming the disadvantages of PARSEC is Sobieczky method for trailing edge modeling. The practical consequence of using this method is a concave surface shaping with curvature increasing towards the trailing edge at both upper and lower surfaces. Such airfoils are known as Divergent Trailing Edge (DTE). This method is mainly based on viscous flow control near the trailing edge that strongly influences aerodynamic efficiency. Fig. 2 illustrates the Sobieczky method for trailing edge modeling. In the simplest case, the parameters  $\Delta \alpha$ ,  $L_1$ ,  $L_2$  that control the increment in trailing edge thickness  $\Delta Z$  are added to make the airfoil surface a divergent trailing edge. The parameter  $\Delta \alpha$  controls the camber added to the upper and

lower surfaces that creates a DTE and  $L_k$  is the chord length measured from trailing edge, which is modified in the Sobieczky method. The function considered for  $\triangle Z$  is:

$$\Delta Z_k = \frac{L_k tan \Delta \alpha}{\mu n} * \left[1 - \mu \xi_k^n - (1 - \xi^n)^{\mu}\right]$$
 (6)

where  $\xi_k$  is the x-coordinate variable. Parameters and variables of this method are illustrated in Fig. 2. The subscript k changes from 1 to 2 in order to consider the length on the upper and lower surfaces, respectively. The shaded region in this figure is the original airfoil generated by the PARSEC method. Different values are possible for parameters  $\mu$  and n (SHAHROKHI y JAHANGIRIAN, 2007).

Figure 4: Sobieczky Method Parameters.



(SHAHROKHI y JAHANGIRIAN, 2007)

0.1.3 Objective Functions. Given a set of design parameters a target function can be defined. This target function must comply with both geometric and dynamic pa-

rameters, hence the importance of the incorporation of accurate physics models. In a CFD-based design process target functions must be evaluated several times until design specifications are met (WU et al., 2007).

To define what an optimum design is after an optimization process, it is necessary to establish the quality of the shape generated. An equation that represents all the criteria defining the characterisites to improve in the design must be introduced. The minimization of this expression (objective function) will lead to an optimum design (FERRANO et al., 2004).

A possible definition of the objective function (OF) is an aggregation function with weighted coefficients  $c_k$  - see (Eq. 0.1.3)-, defining the relevance factors of each single objective  $F_k$ . with K weighting coefficients  $c_k$ , k=1,...,K (GIANNAKOGLOU, 2002).

$$F_{aggr} = \sum_{k=1}^{K} c_k * F_k(x) \tag{7}$$

### (GIANNAKOGLOU, 2002)

In an design optimization process multiple objective functions can be involved, the complexity of the parameterization scales and different approaches must be used. In aerodynamic design it is common to use basis functions that describe the expected behavior of the geometry given specific conditions, then a single optimization function might fail to express the desired objective. Combining with shape parameterization an approach models the geometry as the linear combination of a basis airfoil and a set of perturbation functions, defined either analytically or numerically, These coefficients of the perturbation functions involved are then considered as the design variables. A set of such orthogonal basis functions - see figure 5 - are the functions

to be evaluated to test a design alternative (SONG y KEANE, 2004).

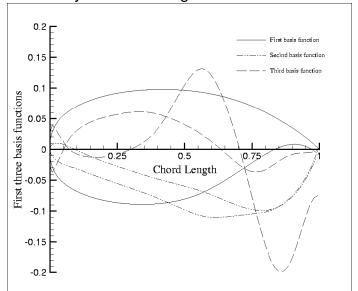


Figure 5: Three numerically derived orthogonal basis functions.

(SONG y KEANE, 2004)

0.1.4 Optimization Methods. The methods available to optimize a given geometry are numerous, and different strategies can be used to find the minimum of the optimization function. We focus on some available and commonly used for airfoil design.

One effective approach to designing wing airfoils with optimal aerodynamic characteristics is that based on the solution of the inverse problems of aerohydrodynamics. The input data are the desired characteristics of the wing airfoil. Multiple parameters like pressure distribution, drag coefficient, lift coefficient -to name a few- are used as bounds for the wing airfoil (ABZALILOV, 2005).

The use of evolutionary algorithms is another strategy commonly used for optimiza-

tion processes. Specifically, the use of Genetic Algorithms (GAs), which are semi-stochastic semi-deterministic optimization methods that are conveniently presented using the metaphor of natural evolution. The GAs are based on the evaluation of a set of solutions, called population. The population is treated with genetic operators: selection, crossover and mutation. All these operations include randomness. The main point is that the probability of survival of new individuals depends on their fitness: the best are kept with a high probability, the worst are rapidly discarded (EPSTEIN y PEIGIN, 2006).

The steepest descent method is an iterative procedure used to accomplish optimization. Starting from some initial geometry the best direction to move from the local point of view is the direction of the steepest descent. This direction corresponds to the negative of the gradient. That is:

$$s^{n+1} = s^n - \beta * \frac{\nabla(f(s^n))}{\|\nabla(f(s^n))\|}$$
 (8)

where  $\beta$  is a constant to be determined (in an optimum way) and the gradient at  $s^n$  of f(s) can be approximated by forward-difference:

$$\frac{\partial f(s)}{\partial s_i}|_{s_n} = \frac{f(s^n + \triangle s_i) - f(s^n)}{\triangle s_i} \tag{9}$$

or with a second order central-difference approximation (HAFTKA y Güdal, 1992):

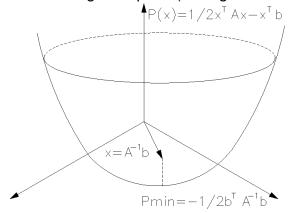
$$\frac{\partial f(s)}{\partial s_i}|_{s_n} = \frac{f(s^n + \Delta s_i) - f(s^n - \Delta s_i)}{2 * \Delta s_i} \tag{10}$$

This method assumes the existence of only one local minimum (unimodal functions).

Under the existence of several valleys the algorithm will converge to the closest local

minimum (see figure 6), which is not necessarily the global minimum. This could or could not be the case and that is why it is important to analyse the general shape of the objective function.

Figure 6: Steepest descent on a given space (local-global minima)



(GARCIA, 2007)

### 0.2 AN INTRODUCTION TO COMPUTATIONAL FLUID DYNAMICS (CFD).

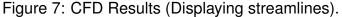
0.2.1 What is CFD. With the continuous development of computer science and hardware, problems which solutions were once almost impossible to obtain or handle due to the large amount of data involved during the process, are now within the reach and capabilities of human beings.

Computational Fluid Dynamics (CFD) is the computer aided practice that deals with the calculation of different flow parameters in a fluid continuum (see Figures 7). The very support of CFD, is none other than the one proposed by Finite Element Analysis (FEA) (See (KIKUCHI, 1986) for an idea of this principles), and as such, the same difficulties that arise in both problem's definition and solution during a FEA analysis,

arise during a CFD analysis.

The first question that has to be asked before starting with a CFD analysis is how the continuum is to be discretize and handled. Mainly, two possible answers might fit the question: to use a mesh dependent method (Eulerian view) or a meshless one (Lagrangian view).

Mesh dependent methods, rely on the fact that a spatial domain  $\Omega$  in  $R^3$  might be discretized small cells to form a volume mesh that approximately represents the original  $\Omega$  domain (KIKUCHI, 1986). Once the discretized domain is set, a set of suitable algorithms is applied to solve the Navier-Stokes equations defined over the domain, obtaining a proper solution to both velocity and pressure variables. The mesh format





(@FUEL TECH INC., 2004)

used to define  $\Omega$  is flexible: can be either regular (hexaedrae) or irregular (tetrahedral), and if the problem is highly dynamic, the grid itself can be modified in time using adaptive mesh refinement methods.

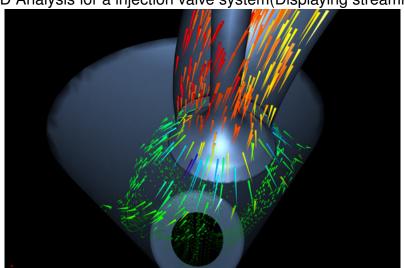


Figure 8: CFD Analysis for a injection valve system(Displaying streamlines).

(@FUEL TECH INC., 2004).

Meshless methods represent a different alternative to the mesh dependent ones: Lattice Boltzmann methods simulating a mesoscopic system on a Cartesian space, Lagrangian viewpoint methods as smoothed particle hydrodynamics and spectral methods (where equations are projected over basis functions such as Chebyshev polynomials) are some methodologies to remark (@WIKIPEDIA, 2007).

As a particular case, it is possible to directly solve set of Navier-Stokes equations for laminar flow cases, as well for turbulent flows in which all relevant length scales are contained within the grid. Other equations, such as the ones related to heat transfer problems or chemical reactions can be solved simultaneously with the Navier-Stokes equations.

More complex CFD algorithms and programs allow the simulation of non-Newtonian fluids as blood and problems including multiphase flows and interaction.

0.2.2 CFD state of the art and current limitants. CFD is undergoing a significant expansion in terms of both number of courses offered at universities and the number of researchers active in the field. There are a number of software packages available that solve fluid flow problems; the market is not as quite as large as the one for structural mechanics codes. The lag can be explained by the fact CFD problems are, in general, more difficult to solve. However, CFD codes are slowly being accepted as design tools by industrial users (FERZIGER y PERIC, 2002).

There are several commercially software packages available, such as FLUENT (@FLUENT INC., 2007), CFX (@ANSYS INC., 2006) and STAR (@ADAPCO INC., 2006), and open source code packages under the GPL license such as OpenFOAM (@OPEN FOAM ORG., 2007). These packages contained within their source codes routines and algorithms to solve incompressible, compressible and multiphase flows, direct numerical and large eddy simulation, combustion and heat transfers problems, but many particular cases and general problems are still impossible to solve with those softwares.

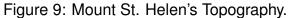
Open source packages as OpenFOAM, even though their learning curve is steep, represent an excellent solution for the dedicated engineer who wishes develop solutions for unsolved problems.

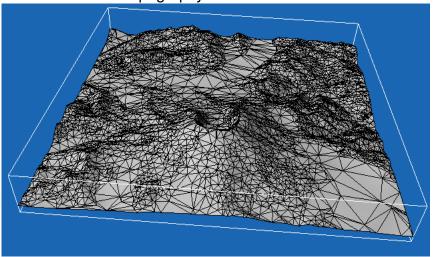
One of the problems faced by CFD developers and users is none other than the one of computational resources. CFD requires high amount of free memory to store up data related to both calculations and results, as well fast processing units are needed to perform the required operations, that could easily ascend to millions per iteration. Parallel processing computing has been the answer to this problematic and right now, big computer clusters are used by the market leaders to suffice their needs.

0.2.3 CFD from the inside, an overview. The CFD method may be subdivided into 3 large steps which cover the whole problem definition and solution process:

### i. Pre-processing.

 Geometry definition. In this stage the physical bounds of the problem are defined. See Figure 9 for a graphical example.

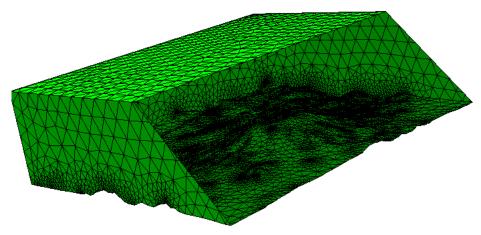




(GARCIA, 2005)

- Geometry Discretization. The volume occupied by the fluid is discretized into finite cells (Figure 10); see DISCRETIZATION METHODS for more information about this topic.
- Physical model definition. Establishing the physical model that will describe the phenomena's behavior; e.g. the Navier-Stokes equations, heat transfer models, entropy equations, etc
- Boundary Conditions definition. After the physical-geometrical boundaries and models are defined, the fluid's behavior and properties at the bound-

Figure 10: Mount St. Helen's CFD domain over the Topography.



(GARCIA, 2005).

aries of the model must be specified; for non-steady state problems (Transient) the initial conditions must be set as well.

- ii. Solution. Once the preprocessing stage is done, a set of equations that define the model is generated given the geometry, boundary conditions and the physical model. This set of equations is solved iteratively (in most of the cases) for both steady state and transient problems. Diverse strategies are used to solve the huge equation systems involved during this step, such as Gauss Seidel, Successive overrelaxation or Krylov subspace methods (BARRET, 1994).
- iii. Post-processing. Post-processing is the final step of the CFD process. After solving the problem, visualizing the solution is an important matter that is vital to formulate a proper analysis and solution validation judgments. Stereoscopic view systems, high definition screens and immersive virtual reality environments are used for this purpose on high end systems and large scale projects.

0.2.4 Discretization Schemes. The selection of a suitable discretization scheme to solve a problem is one of the keys to success. For several problems, more than one discretization scheme is appropriate and right on this spot is where the criterion of the person who's facing the situation is definitive (solution times might vary from scheme to scheme exponentially for a given problem). Numerical stability is one of the main goals one must pursue, and in most of the cases the discretization scheme chosen is the one that fits best this condition. Some of the most used discretization methods being used are:

Finite Volumes Method. This method responds to the principle that the governing equations for the domain can be solved on discrete control volumes that represent the original domain; The integral approach for this method can be seen on the following equation:

$$\frac{\partial}{\partial t} \int Q \, dV + \int F \, dA = 0 \tag{11}$$

Where Q is the vector of conserved variables, dV is a differential of Volume, F is the vector of fluxes and dA is the boundary of each differential of volume. It is evident that this method is inherently conservative. This is a very popular methodology on both commercial and open source codes.

Finite difference method. This method has historical importance and is simple to program. It is currently only used in few specialized codes. The main disadvantage is that it requires structured meshes, and coordinate transformations for complicated geometries (@WIKIPEDIA, 2007).

Finite Elements Method. The development of this methodology can be tracked back to the 1940's. The main idea of this approach is the possibility to divide

a continuous domain using a mesh discretization, returning a set of discrete sub domains. (See 11) The method was provided with a rigorous mathematical foundation in 1973 with the publication of Strang and Fix's *An Analysis of The Finite Element Method*, and has since been generalized into a branch of applied mathematics for numerical modeling of physical systems in a wide variety of engineering disciplines, e.g., electromagnetics and fluid dynamics (@WIKIPEDIA, 2006b). Although this methodology is used mainly for structural and mechanical analysis, it is also used for fluids; However, this formulation requires a special care for transient situations to ensure conservative solutions.

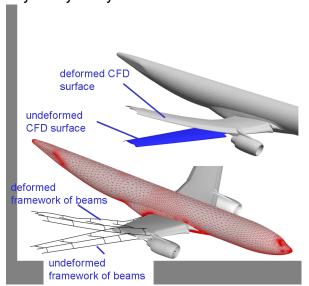


Figure 11: Aeroelasticity analysis by FEM.

(@CATSWEB, 2007).

Boundary Element Method. Conceptually, this formulation works by constructing a mesh over the modeled boundary. This formulation has become more popular since the 1980's. Because it requires calculating only boundary values, rather than values throughout the space defined by a partial differential equation. It

is significantly more efficient in terms of computational resources for problems where there is a small surface/volume ratio (@WIKIPEDIA, 2006a).

- 0.2.5 CFD flow behavior models. Most flows encountered in engineering practice are turbulent and therefore require different treatment. Turbulent flows are characterized by the following properties (FERZIGER y PERIC, 2002):
  - Turbulent flows are highly unsteady. A plot of the velocity as a function of time at most points in the flow would appear random to an observer unfamiliar with these flows. The word 'chaotic' could be used but it has been given another definition in recent years.
  - They are three-dimensional. The time-averaged velocity may be a function of only two coordinates, but the instantaneous field fluctuates rapidly in all three spatial dimensions.
  - They contain a great deal of vorticity. Indeed, vortex stretching is one of the principal mechanisms by which the intensity of turbulence is increased.
  - Turbulence increases the rate at which conserved quantities are stirred. Stirring
    is a process in which parcels of fluid with differing concentrations of at least
    one of the conserved properties are brought into contact. The actual mixing is
    accomplished by diffusion. Nonetheless, this process is often called turbulent
    diffusion.
  - By means of the processes just mentioned, turbulence brings fluids of differing momentum content into contact. The reduction of the velocity gradients due to the action of viscosity reduces the kinetic energy of the flow; in other words,

mixing is a dissipative process. The lost energy is irreversibly converted into internal energy of the fluid.

- It has been shown in recent years that turbulent flows contain coherent structuresrepeatable and essentially deterministic events that are responsible for a large
  part of the mixing. However, the random component of turbulent flows causes
  these events to differ from each other in size, strength, and time interval between occurrences, making study of them very difficult.
- Turbulent flows fluctuate on a broad range of length and time scales. This
  property makes direct numerical simulation of turbulent flows very difficult.

#### 0.3 AN INTRODUCTION TO WING AERODYNAMICS

Aerodynamics is an applied science dedicated to the study of the behavior of the air when a body moves through it or a body opposes to its flow. One of the most important manifestations of the interaction between fluid and solids is pressure. Pressure is a measurable change of energy of the fluid molecules colliding with the surface of the solid. For each point of the surface a pressure magnitude can be measured (ANDERSON, 2001). In general efforts on the study of aerodynamics aim to:

- Prediction of forces and moments on, and heat transfer to, bodies moving through a fluid. i.e. Lift, Drag, Moments, Heating of airfoil, airplanes, buildings, etc.
- Determination of flows moving through ducts. i.e. Calculation of air-breathing jet engines, engine thrust, etc.

- 0.3.1 Aerodynamic Variables. The variables necessary to deal with an aerodynamic study are described below (ANDERSON, 2001):
  - Pressure is the normal force per unit area exerted on a surface due to the change of momentum of gas molecules impacting the solid.
  - Density is the mass per unit volume
  - Temperature is the measure of kinetic energy of the molecules of the moving fluid.
  - Viscosity
  - Reynolds number (Re) is the physical measure of the ratio of inertia forces to viscous forces in a flow.

$$Re = \frac{\rho * v * l}{\mu}$$

 $\rho:\ density$ 

v: velocity.

 $\mu$ : viscosity.

 $l:\ characterisitc\ length.$ 

• Mach number is the ratio of the flow velocity to the speed of sound.

$$M = \frac{V_{\infty}}{a_{\infty}}$$

 $V_{\infty}$ : freestream velocity

 $a_{\infty}$ : freestream area.

Table 1: Relation between Mach number and flow regime

Flow regime	Mach Number
Subsonic	M < 0,7
Transonic	0.7 < M < 1.2
Supersonic	1.2 < M < 5
Hypersonic	M > 5

(ANDERSON, 2001).

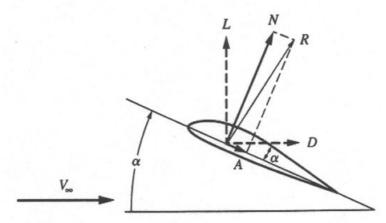
The understanding of these variables allow correct interpretations of the phenomena involved in fluid analysis.

0.3.2 Aerodynamic Forces and Moments. Aerodynamic forces are the reactions of a body moving through a fluid. Pressure and shear stress distributions on the body surface are the acting forces that make flight possible. Different force scenarios can be observed depending on the orientation of the body. Pressure (p) acts in the normal direction to the body and shear  $(\tau)$  acts tangentially (see figure 12). The effect of p and  $\tau$  distributions integrated over the complete body surface is a resultant aerodynamic force R and moment M on the body. R can be split into two components lift (L) and drag (D) (ANDERSON, 2001).

0.3.3 Wing Profiles. In aeronautics a wing profile or foil is a curved shaped contour able to generate lift by generation of a distributed pressure on its surface. This specific geometry in aerospace studies has the following typical regions (see figure 13):

- Leading Edge (LE).
- Trailing Edge (TE)

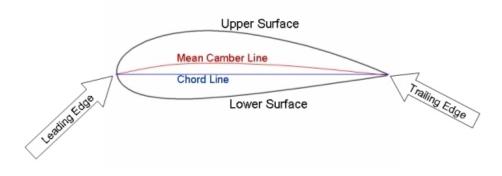
Figure 12: Wing airfoil aerodynamic forces decomposition.



(@AEROSPACEWEB, 2007).

- Upper Surface (US)
- Lower Surface (LS)

Figure 13: Wing airfoil typical regions.



(@DREESECODE, 2007).

The geometrical relations between the defined regions have specific names. Depending on the definition of these parameters the aerodynamic behavior of the foil will be altered. Experts use other important characterisites for the design and analysis of airfoils. In figure 13, the chord line is presented as the linear distance from

the leading edge to the trailing edge. The airfoil thickness is the maximum distance between the upper and lower surfaces, and the attack angle previously defined as the angle of incidence between the chord line projection (see figure 12) and the flow (ANDERSON, 2001).

The Lift Coefficient ( $C_i$ ) relates the lift force (L) with body shape and fluid properties.

$$C_{l} = \frac{2 * L}{\rho * V^{2} * S}$$

$$L = Lift.$$

$$\rho = Density$$

$$V = Velocity.$$

$$S = Reference area$$

The drag coefficient  $(C_d)$  relates drag force (D) with the shape of the wing and the fluid properties.

$$C_d = \frac{2 * D}{\rho * V^2 * S}$$

$$D = Drag.$$

$$\rho = Density$$

$$V = Velocity.$$

$$S = Reference area$$

The moment coefficient  $(C_m)$  relates the twisting moment (M) to a point, usually this point is set to 1/4 of the mean chord from the leading edge.

$$C_m = \frac{2 * M}{\rho * V^2 * S}$$
 $M = Moment.$ 
 $\rho = Density$ 
 $V = Velocity.$ 
 $S = Reference area$ 

The defined coefficients determine the correct behavior of wing profiles. An appropriate combination of these variables for a given attack angle and speed yield smoother results.

#### 0.4 OPTIMIZATION IN COMPUTATIONAL FLUID DYNAMICS

CFD has come a long way since its beginings. The speed and capacity of the hardware to manipulate huge sets of data has increased considerably. This capacity allows to simulate complex fluid flow situations. Problems that were previously addressed with different disciplines are now approachable taking advantage of CFD. The robustness offered by current CFD codes to solve the governing flow equations, enable designers to obtain accurate preliminar field data to estimate the response of new objects from virtual prototypes. One of the advantages of using CFD analysis for optimization is to have multiple sets of data available at the same time (volumetric fields). The possibility to analyze transient flow cases gives a head start to understand the behavior of object in more realistic scenarios. Multiple efforts have been

carried out regarding coupling of CFD analysis and design optimization, several inhouse codes have been developed as well as existing cad tools and code integration.

As technological improvement and competition require more careful optimization of designs or, when new high-technology applications demand prediction of flows for which the database is insufficient, experimental development may be too costly and/or timeconsuming. Optimization in these areas can produce large savings in equipment and energy costs and in reduction of environmental pollution (FERZIGER y PERIC, 2002).

The calculations obtained from CFD can be used for optimization regardless of the complete accuracy of the model. The predictions obtained when turbulence models are used are not accurate enough that they can be accepted quantitatively without testing. However, the trends may be accurately reproduced so that the design predicted to be the best by the model also performs the best in tests. Calculations based on turbulence models can reduce the number of experimental tests required and thus reduce the cost and the time required for development of a new product (FERZIGER y PERIC, 2002).

The use of k-epsilon turbulence model is quite popular, although it has been known that there is a deficiency in its performance for problems involving rotation and curvature, the standard k-epsilon turbulence model is used in widely in studies for the steady-state turbulent flow calculations, due to its robustness in practical applications (WU et al., 2007).

Multidisclipinary Optimization (MDO) presents optimization on a new perspective.

MDO integrates CAD softwares to control design parameters and CFD simulation softwares to acquire the data to evaluate objective functions (DYE et al., 2007). Optimization then has a physics based data to evaluate design options. The modeling of gas turbines are among the most complex systems available, a cad based parametric approach has rendered interesting results using the following methodology (see Fig.14).

Customer Requirements and Vehicle Model Characteristics
• Flight Conditions • Bleed Flows
• Compressor Design Point Pressure Ratios • Inlet Flows Customer Requirements · Component Design Point Efficiencies Duct Losses · Thrust or T4 **NPSS** NASA NPSS (Numerical Propulsion System Simulation) output parameters Station† Temperatures & Pressures (initial estimates) Diameters at Stations Propulsion Pressure Ratio or Work Requirements System Init al Non-dimensional Gaspath Representation Parameters Mach Numbers
 Aspect Ratios Hub/Tip Ratios Specific Flow
 Airfoil Counts Flowpath Rotor Speeds Geometric & Cycle output Flowpath Cycle Output Parameters Refined Station† Temperatures
 Refined Station Pressures parameters Flow performance characteristics Airflow
 Speed lines CSD Surge Margin
 Refined, Dimensional Geometric Gaspath Representation **CAD Assembly Parameters**  Axial & Radial Cornerpoint locations
 ID and OD Endwall slopes CAD Model · Airfoil parameters: Chord, Span, AR, Gap, Cant Angle CAD · Airfoil Counts Assembly † Stations: Fan Inlet/Exit, Compressor Inlet/Exit, Combustor Inlet/Exit, Turbine Inlet/Exit, Mid-module locations

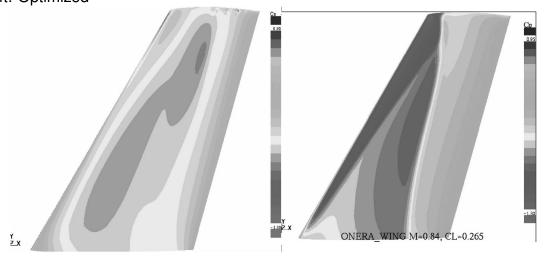
Figure 14: Engine Cylce Modeling Sysem Components and Input/Outputs

(DYE et al., 2007).

Some other approaches include the development of the CFD solver for the Navier-Stokes equations governing the proposed type fo flow. The task of programming a solver offers more suitability for optimization. The optimization involves high-dimensional search spaces, and the non-triviality of the search for the optimum increments the computational cost. The numerical solution of the full Navier-Stokes equa-

tions can be based on a multiblock code NES that employs structured point-to-point matched grids. The key feature of NES is the use of the Essentially Non Oscilatory (ENO) numerical scheme. The ENO approach is a high-order approximation scheme designed for solutions containing discontinuities. In Navier-Stokes computations, the scheme is usually applied to the approximation of convective terms (EPSTEIN y PEI-GIN, 2006).

Figure 15: CFD-Based Optimization, pressure distribution. Right: Original Wing, Left: Optimized



(EPSTEIN y PEIGIN, 2006).

A computational fluid dynamics-based design system with the integration of three blade design approaches, automatic mesh generator and CFD codes enables a quick and efficient design optimization of turbine components. This examples include sophisticated Large Eddy Simulations (LES) in a Francis turbine and in a centrifugal pump impeller at design and off-design conditions. However, a robust and fully 3D inverse design approach, by which the required flow characteristics and parameters are specified as inputs and the corresponding blade geometry is computed and generated as output, is still not commonly implemented. The governing equations for

this phenomena are for turbulent flow, but the current assumptions still dwell on the inviscid approach. A viscous CFD solver is needed. The design has to be evaluated by the solver and the solution must be updated to modify the input (WU et al., 2007). The inviscid Q3D codes by means of both finite difference method and FEM incorporated into this system are primarily employed in the preliminary optimization stages due to their rapid convergence rate and reliability. Using the following methodology (see Fig.16) a francis tubine is optimized. Developed for nearly 2 decades, mesh handling and generation, CFD analysis and design optimization is integrated under one single iterative process. As an example they obtain this result in redesigning the

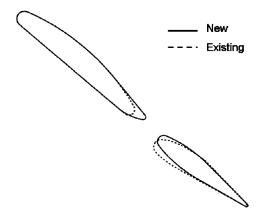
**CFD-Based CAD Design System** H, Q, P, n **Preliminary Design by Blade Design Methods Grid Generation** Geometry Rapid CFD Modification Q3D or Euler no O.K? yes **Elaborated CFD** 3D Visualization K-€ or LES <u>......</u> **Stress Analysis Model Test** no Manufacture O.K?

Figure 16: CFD-Based Design System

(WU et al., 2007).

vanes of the runner (see Fig. 17).

Figure 17: Optimized guide vane and stay vane profiles



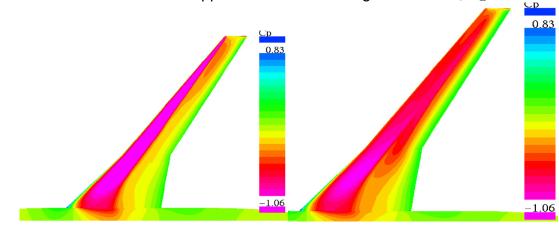
(WU et al., 2007).

There are also fully 3D approaches on CFD optimization under development. Evaluating 3d lifting surfaces for wing-body aircraft configurations is one of them. By fixating the lift, minimization of drag is sought given numerous geometric and aerodynamical constraints. The optimization method is based on the use of Genetic Algorithms, accurate full Navier-Stokes drag prediction and massive multilevel parallelization of the whole computational framework. This methods highlights are multipoint wing optimization for transport type aircraft configurations (see Fig. 18). Reducing in the drag even in 1% would yield noticeable results in th pay-load of the aircraft (EPSTEIN y PEIGIN, 2007). It was proven that this method allos to design feasible aerodynamic shapes which:

- Possess a low drag at cruise conditions;
- Satisfy a large number of geometrical and aerodynamic constraints (15-20 per design);
- Offer a good off-design performance in markedly different flight conditions such

as take-off and high Mach zone.

Figure 18: Original(Left) and Optimized (Right) ARA M-100 wing-body configuration. Pressure distribution on the upper surface of the wing at M = 0.80,  $C_L$  = 0.50.



(EPSTEIN y PEIGIN, 2007).

#### 1 METHODOLOGY

The main objective is to produce a method that optimizes the shape of an aerodynamic profile. To achieve this, we must define first the set of needs towards the construction of the method itself. The process begins with the identification of the disciplines to integrate:

- i. Shape parameterization.
- ii. Geometry manipulation.
- iii. Aerodynamics.
- iv. CFD-Based shape optimization.
- v. Optimization methods (Gradient-based).

The steps to follow become the sewing thread between the mentioned disciplines. They conform a road map of the whole process but it is not required to stay in sequential order. The basic scheme is presented as follows:

- Initially, a geometry is defined as the input and target of the method.
- The set of design variables are defined (they are a requirement of the method).
- Definition of the optimization functions.
- The geometry has to be represented as a data set that is portable between the multiple tools.
- Parameterize the geometric characteristics of the body to allow portability and increase simplicity-.

- The geometry has to be prepared for CFD analysis, inlcuing generation of the domain and boundary conditions.
- Adaptation of a chosen CFD solver to allow evaluation of optimization functions.
- Perform a CFD analysis on the given geometry, including embedded evaluation of the optimization functions during run time.
- Manipulate the geometry based on the optimization method and criteria.

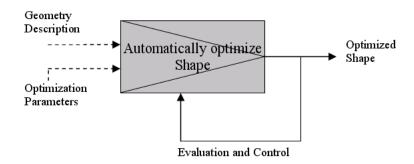
In general, if the process is contemplated as a black box it can be simplified to basic input/output variables. The main I/O variables will begin the definition of the structure of the method.

During any design process, feedback is an important step before any design alternative is chosen. The expertise of the designer can accelerate an optimization process avoiding stagnation in local minima. This reason supports the need of offering a degree of interaction between the optimization process and the user. Finding optimal solutions with the minimal effort is foremost the objective to achieve, and reducing the problems to its general form help to address their solution. Basically, the set of tasks of the process can be reduced to the following needs:

- An application that integrates a set of tools to test and analyze automatically an aerodynamic profile.
- A set of tools to perform the required tasks for the optimization process.

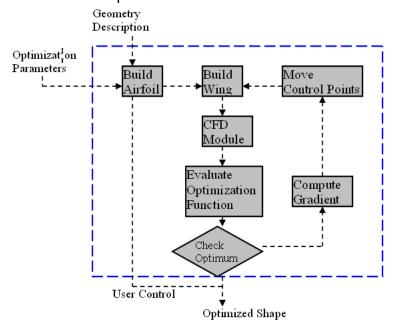
Then a black box representation like in Fig. 1 initiates the design process. Further segregation proves itself useful in the definition of the tasks to perform. Placing the problem in a form of structure before contemplating any data types (or specific solution methods) can bring forth new sets of solutions. This strategy also reveals the

Figure 1: Black Box Representation of the Method



subdivision of functions and the expected interaction between them.

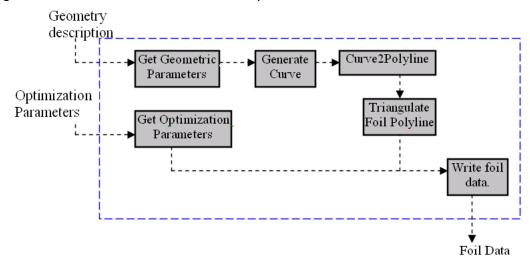
Figure 2: Function decomposition



In Fig 2. an overall decomposition in basic sub-modules can be appreciated. From the conception of the input data, to the generation of the base airfoil, up to its analysis under a CFD environment and the manipulation of its geometry, the tools that must be developed to solve each portion of the problem arise clearly.

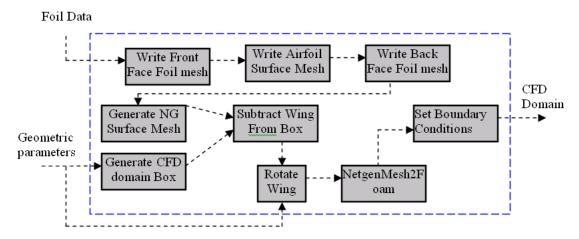
In Fig 3 the functions to generate an airfoil from basic geometric data is presented. The input file represents the type of parameterization used, since the definition of the

Figure 3: Build Airfoil Function Decomposition



control points defining the upper and lower surfaces of the arifoil profile are needed.

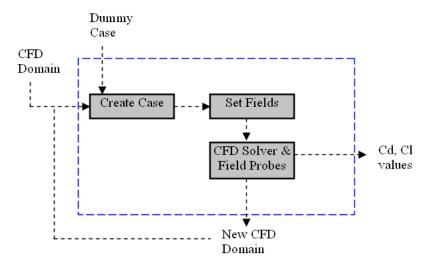
Figure 4: Build Wing Function Decomposition



Interaction between modules starts to show, when data obtained as a 2D airfoil has to be translated to a 3d wing, here previously acquired geometric data come into play. In order to obtain a feasible simulation meshing processes have to be coupled with CFD domain generation. The correct balancing of the parameters play an important role in the generation of a valid CFD domain.

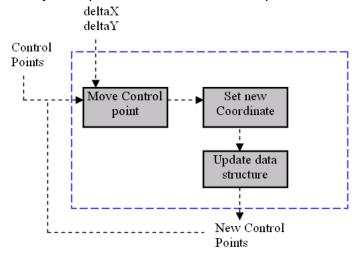
Instead of undertaking the task of developing a CFD solver, some open source codes

Figure 5: CFD module Decomposition



like OpenFOAM offer a variety of solvers and tools for retrieving data. The structure of such software and its Object Oriented Programming (OOP) nature enables the user to implement a customized CFD solver up to his needs. In the present method an adaptation of a turbulent steady-state solver using SIMPLE algorithm was implemented. It was further modified to allow reading of field to calculate drag and lift coefficients.

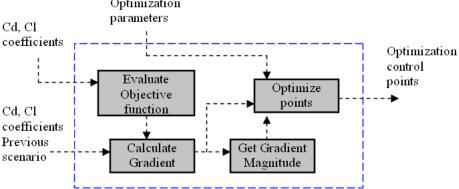
Figure 6: Geometry Manipulation Function Decomposition



In order to achieve an optimal shape, modifications of the geometry have to be done to estimate the possible change in shape. The manipulation of the geometry can increase in difficulty as the number of parameters and control points rise. Increased numbers of control points will generate bigger search spaces, since each control point perturbed means a complete CFD simulation with the correspondent objective function evaluation.

Figure 7: Shape Optimization Function Decomposition

Optimization



#### 2 FOIL SHAPE OPTIMIZATION APPLICATION DESCRIPTION

#### 2.1 Presentation.

In this chapter, the user will find how to install the developed tool for airfoil shape optimization, where the components are located and how to compile them. This chapter consists of basic notes about Netgen (NG) installing and it's requirements, triangle code compilation, b-Curves library inclusion and OpenFOAM (OF) integration.

The foilOptimization application can be catalogued as an automatic CFD-based optimization solver for incompressible turbulente Navier-Stokes equations using the SIMPLE algorithm. The original simpleFoam solver is adapted to allow drag and lift readings during run time. This code is inehrently steady state, condition considered to be accurate for the problem setup.

The foilOptimization application depends on a set of open source packages to perform the different operations required to manipulate the geometry, solve the CFD problem and apply an optimization technique. In the following sections, each of the applications will be briefly described at two levels: how to install it and how does it interacts with the application.

#### 2.2 Triangle

Inside the foilOptimization application, triangle is the tool that takes care of the 2D triangulation of the polyline describing the airfoil. The data structures of triangle are used throughout the code to preserve the definition of the geometry, and allow regeneration of the airfoil polyline. For more information see SHEWCHUK (2007).

### 2.3 B-Curves

The role of B-Curves in the application is to provide the tools to manipulate B-Spline curves for the parameterization of the profile geometry. B-Curves is a library programmed within the frame of the Applied Mechanics lab, it provides the underlying structures for the manipulation of teh geometry. In combination with triangle they represent a CAD emulation.

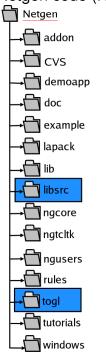
# 2.4 Netgen.

Inside the foilOptimization application, NG is the one who deals with the proper acquisition of the wing's boundary representation (B-REP) and in a later stage, the generation of a valid triangular surface mesh to be used as the inner object representation inside the CFD scenario.

The object's B-REP is acquired from a combination of the integration of the triangle code and own-developed tools to generate a 3d wing from a 2d profile. Currently, valid ASCII Stereolitography (STL) are used for geometry acquisition, providing valid engineering tolerances for design process.

The current version of NG source code used for this purpose is v4.5. Only two large parts of the source code are used in this application: the files related to the meshing and geometry acquisition algorithms (located inside the libsor directory) and the files related to the NG GUI widgets (located inside the togl directory, this files are maintained due to compilation issues). (See Figure 1).

Figure 1: Required portions of the Netgen code (Highlighted)



This two directories of the NG source code are copied and translated into the root directory that contains all the foilOptimization source code, as shown in Figure 2.

IMPORTANT: While compiling Netgen 4.5 source code, a proper installation of Tcl/Tk, OpenGl and Tix is needed. Beware that the availability of Tcl/Tk source code is a MUST to compile NG and its sources. See SCHOBERL (2006b),SCHOBERL (2006a),SCHOBERL (2005) and NG README file for more information.

Figure 2: Location and distribution of the source code of Netgen, BCurves, triangle and foilOptimization.



#### 2.5 Data flow review.

Given a geometry and design variables defined by an user, this application optimizes the shape of an airfoil. The input data file allows some degree of interactivity since it allows to stop the process at any stage. The selection of the parameters sometimes must be done intuitively. The input file is the option of interaction between the code and the user to iterate over this parameters.

The inputs required are the control points, triangulation parameters, meshing parameters, optimization parameters, and 3d geometry data. The increased number of inputs is a consequence of generalization of the application; for each geometry (and depending on its complexity) different meshing and optimization parameters are needed, otherwise, teh method will not converge.

## 2.6 Current application compilation.

The root directory of the fiolOptimization application source code is currently to be located on the same directory where all the OF solvers for incompressible flow are positioned (simpleFoam, boundaryFoam, nonNewtonianIcoFoam, etc...). This directory can be reached (after a proper OF installation and setting of system variables) at \$FOAM\_APP/solvers/incompressible.

The main directory is called foilOptimization, it contains a basic makefile (Makefile) that compiles the application's library (libfoilOptim) that contains all the NG, Bcurves, Triangle and other objects of the foilOptimization application, to be linked later with the other OF libraries using OpenFOAM's own wmake. All the required files for wmake compilation of the application are located at:

\$FOAM\_APP/solvers/incompressible/foilOptimization/Make

. For more information about how does wmake works, see WMA (2000-2007).

To compile the application, just type in the command line (while being in the \$FOAM\_APP/solver directory) make. The application's executable will be placed on the same directory, under the name foilOptimization.

# 2.7 Current foilOptimization application code Hints.

The foilOptimization source code lies mainly on 5 files, with the following description:

foilOptimization.C: This file contains the <u>main</u> routine of the application and it's
the starting point of it. Here all the pieces of code meet, a set of functions are
defined in this file regarding CFD solving and CFD domain generation.

- CFD\_DomainMesh.cpp: This file contains all the functions to manipulate the CFD domain generated using netgen.
- CFD\_DomainMesh.H: Contains all the defines (Classes, constants and macros required in the CFD\_DomainMesh.cpp file).
- Airfoil.cpp: Contains the methods to manipulate, construct and evaluate teh geometry under the parameters of the optimization, it integrates BCurves.C and Triangle.C to allow geometry generation and manipulation.
- Airfoil.h: It contains the defines, class methods, constants needed by Airfoil.cpp.
- 2.7.1 OpenFOAM patch detection over unstructured CFD doamin mesh. During the construction of a fvMesh, the user must supply to OF a valid list of indexed boundary faces. Having explicit information about those boundary faces (Do they belong to the boundary with the object? or, Do they belong to the CFD Domain boundary?).

After performing the generation of the *fvMesh* given the cell data (vertex, cell types and connectivity), OF catalogues all the un-neighbored faces of each cell by default as defaultFaces, and storages each face index into a list. As a post-condition, all the boundary faces of the CFD doamin are stored into a single list, which only has information about indexes.

Extracting the boundary of the CFD domain is still insufficient to perform the CFD analysis: It is still needed information about where to establish the boundary conditions on the geometry. OF uses a figure, called patch to map the desired boundary conditions over grouped set of faces.

2.7.2 Boundary condition setting over unstructured CFD domain mesh. After filtering the OF patches, boundary conditions are to be set over them. The correspondant boundary conditions, as defined by OF, should be as follows:

i. Object Boundary: Fixed Walls.

ii. Inlet: Inlet (Velocity has to be set -use a dummyCase-).

iii. Outlet: Outlet (Fixed Pressure value).

iv. CFD domain Walls: symmetryPlanes.

### 3 TESTS AND RESULTS

The algorithm was tested using 6 different geometries, 3 of them were NACA profiles. The results observed in the NACA standard pfofiles are encouraging, even thoough these profiles are already tested. The other 3 geometries are just proof of concept's examples, they help to expand the understanding of the job done by the algorithm.

It is to be expected form aerodynamic principles that a decrease in drag will carry on a decrease in lift. The objective is to achieve a decrease of drag with the minimal lift loss possible.

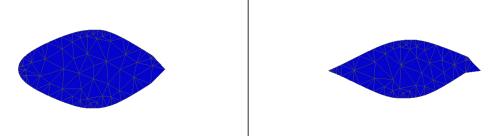
#### 3.1 Overall review of reuslts

Table 3.1: Comparisson between coefficients

Example Name	Cd	Cd new	% diff	CI	CI new	% diff
NACA0012	0.0181661	0.000994103	94.5	0.00326688	0.000283193	91.3
NACA0022	0.0210092	0.00089619	95.7	0.0107231	0.0001874	98.25
NACA0030	0.0556766	0.00539162	90.3	0.0187379	-5.34506e-05	99.7
airfoil	0.254807	0.0187984	92.6	0.00036265	0.00539171	-48.67
airfoil_raw	0.479625	0.0519777	89.2	0.0176995	0.0105523	40.3
cylinder	0.444139	0.0181473	95.9	-0.0283842	-0.00119709	95.7

In table 3.1 we can observe clearly the changes experienced by the geometry; moreover the geometries that were considered for the proof of concept experienced the most significant changes. In Fig. 1 the geometry was considerably altered to reduce drag. The resulting geometry will blend with the surrounding flow in a better way.

Figure 1: Shape optimization of an example arbitrary profile



# 3.2 Optimization of NACA airfoils

In the set of NACA airfoils, the contour was obatined from (TRAPP y ZORES, 2007). This Java applet generates the x and y coordinates for many 4 digit NACA foils. The minimum number of controil points to describe the airfoil were used. Having a reduced set of control points reduces the search space, then the whole run for the application takes less time. Remember that for each control point perturbed a complete CFD simulation case is generated (see Fig. 2).

Figure 2: Case structure of an optimization



The set of NACA foils were not altered as much as teh other examples, but the slightest variation to this standard and tested profiles proved to change their coefficients considerably. In Fig. 3 the change in shape may not be noticeable, but when teh results from the coefficients are read, the difference can be seen. A considerable change is also noticeable in the pressure distribution see Fig. 4 and note the data bars. This type of visualization can be obtained when using full CFD approaches. Each CFD simulation holds field data useful for multiple analysis beyond the objective functions only. The change in he performance of the whole wing rather than the in the mesh shape only is observed.

Figure 3: NACA0030 optimization

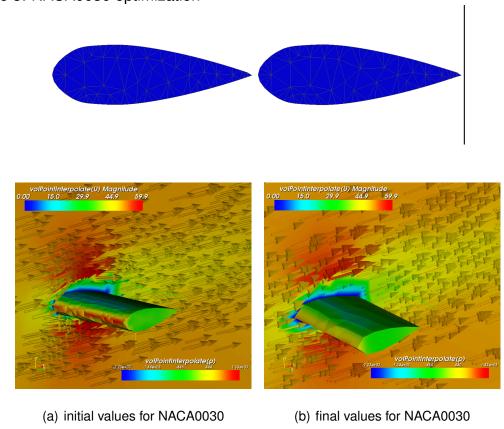
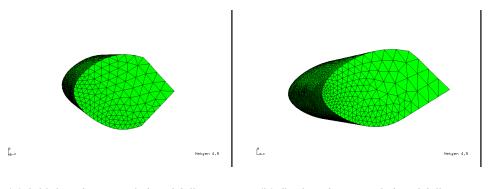


Figure 4: NACA0030 optimization CFD visualization

# 3.3 Conceptual Examples

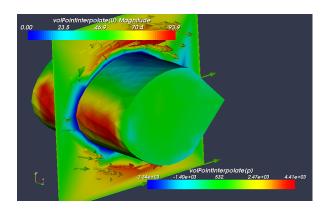
The final example is the most representative in terms of change of shape, nonetheless it is not an standard airfoil. It fulfills the requirement of proving the concept. With just 2 control points in the upper surface and 2 control points on the lower surface (given that the extremes remain fixed) a noticeable and encouraging change in shape was observed.



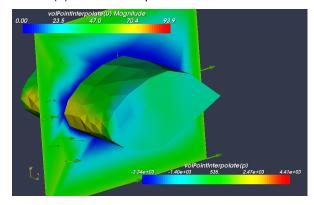
(a) initial surface mesh for airfoil\_raw

(b) final surface mesh for airfoil\_raw

Figure 5: Representative airfoil optimization



(a) initial u and p fields for airfoil\_raw



(b) final u and p fields for airfoil\_raw

Figure 6: CFD visualization of the airfoil\_raw case

### 4 CONCLUSIONS AND FUTURE WORK

#### 4.1 Conclusions.

amounts of disk space.

A method for the CFD analysis and simulation of aerodynamic profiles is presented on this work. The proven concept was to optimize the shape of an aerodynamic profile using a gradient-based method. After intensively selecting the optimization criteria and constants a satisfactory condition was achieved. When pursuing an objective like reducing drag, lift is proportionally reduced as well since they are directly related.

An optimal combination of forces is the objective of the method. An accurate definition of the constants that define the weight of the objectives becomes a crucial step. When only a minimization of drag is trated, teh geometry differs greatly from the initial, but the significant loss of lift leads to profiles with no capacity to sustain flight conditions optimally. For this matter, a penalization of lift and minimal area acceptable were introduced.

The possibility of exploring drag minimization exclusively in aerodynamic design poses interesting results. Using the developed method with switches for specific design scenarios (automotive, hydrodynamic) offers feasible and impoved shapes. One of the main concerns during the tests was the quality of the mesh. The high curvature involved in the leading edges of airfoils forced the need for incresed number of surface elements, this translated into elevated number of tetrahedra in the CFD domain volumetric mesh, inducing high demands on computational operations and

Coarser meshes allow the designer to have valuable initial approximations. If an accurate mesh grade is defined, even if coarse, the designer can obtain shape op-

timization within minutes. The ability of the code to define a shape with few control points presents and important advantage when the optimization process occurs. The dimension of the search plays a vital role when defining the evolution of the shape, having fewer control points reduces the size of the search space, enabling the designer to obtain intial approximations faster.

A limitant when analyzing different shapes becomes the particular boundary conditions of each shape for its CFD analysis, an interactive process to select this conditions would prove itself useful. Setting the conditions of the turbulent solver are higly case-dependent as well. In this work the k-epsilon model was used due to its robustness in practical application. When high Reynolds numbers lead to the assumption of a steady state flow condition, the k-epsilon model is used widely.

The process of constructing the shape in 3d from 2d control points and later on the CFD domain became a time consuming task. A parametrical definition of the geometry was implemented and allows the generation of various shapes with a valid CFD domain. The CFD domain size is quite large when treating outside flows (like the ones when simulating an arifoil moving through air) and causes simulation times to increase considerably.

### 4.2 Future Work.

After proving the concept of CFD-based shape optimization in a single CPU, the code will undergo changes to allow parallelization. Parallelization will allow more complex geometries, more refined domains and shorter simulation times.

The current meshing process has to be reviewed for efficient adaptation to CFD optimal conditions, possible use of more specific airfoil hexahedra mesh generation will be explored using OpenFOAM's embedded mesher "blockMesh".

Exploring the integration with a CAD platform under Linux (the native environment of the application) in C/C++ programming language would scale the usability of this application. Currently the API of ProEngineer offers an alternative to such challenge. Further study in optimization objective functions is yet to be done. Work has but begun....

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