



Aerodynamic Optimisation Using CAD Parameterisations in SU2

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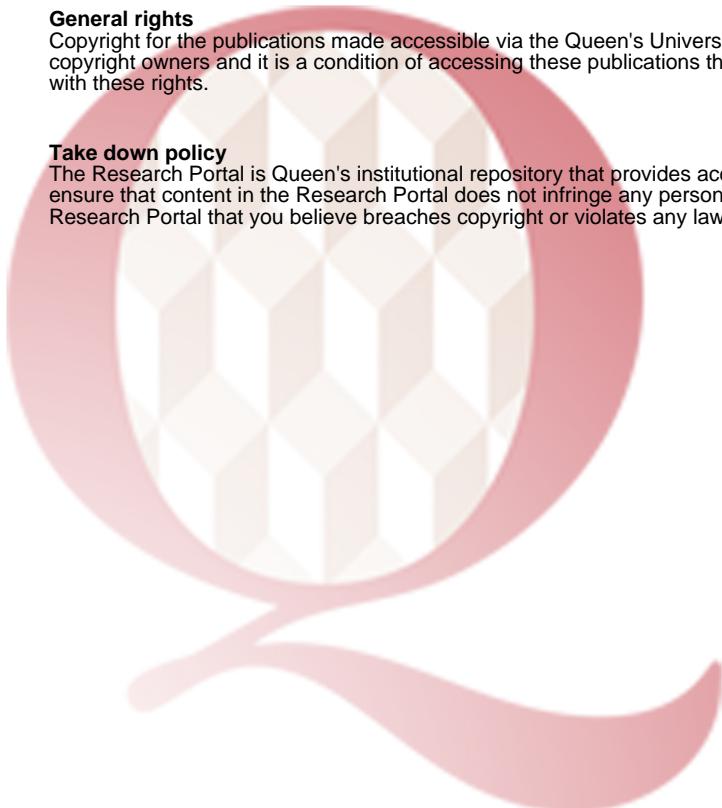
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Aerodynamic Optimisation Using CAD Parameterisations in SU^2

4th EASN Flight Physics Workshop
Aachen, Germany

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Outline

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1 Overview

2 Motivation

3 Gradient Calculation

4 Test Cases

5 Conclusions

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- Results
- Conclusions

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3 Gradient Calculation

4 Test Cases

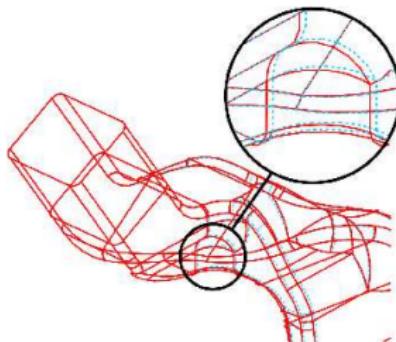
5 Conclusions

Motivation

- Increase flexibility of Adjoint Based Optimisation

Motivation

- Increase flexibility of Adjoint Based Optimisation
- Efficient calculation of any parametric sensitivity

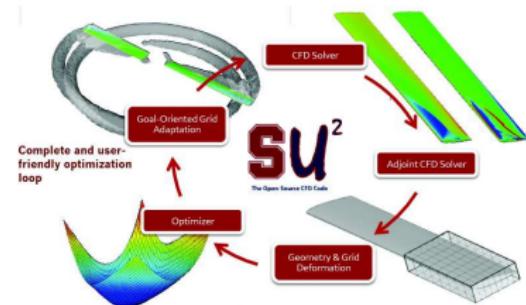


SU²

SU² is an open-source CFD/Adjoint optimisation framework¹

- Developed at Stanford University

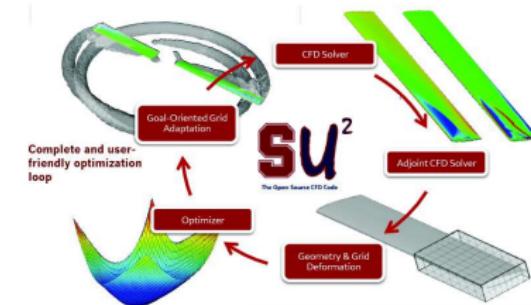
SU²
The Open-Source CFD Code



SU²

SU² is an open-source CFD/Adjoint optimisation framework¹

- Developed at Stanford University
- General purpose PDE solution methods



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- Range of numerical schemes available (JST, ROE, MG, Euler-Implicit,...)



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SU² is an open-source CFD/Adjoint optimisation framework¹

- Developed at Stanford University
- General purpose PDE solution methods
- Range of numerical schemes available (JST, ROE, MG, Euler-Implicit,...)
- Mesh deformation/adaptation

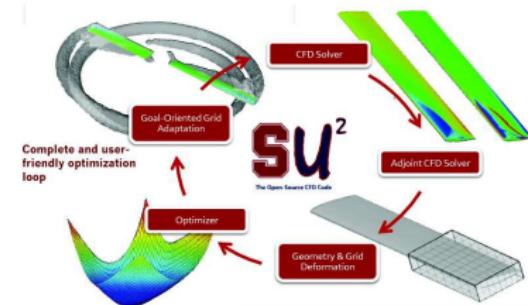


SU²
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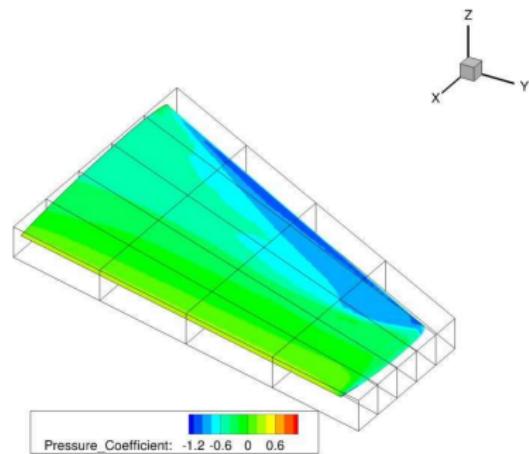
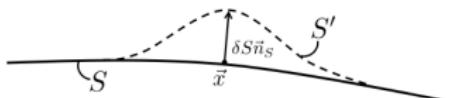
SU²

SU² is an open-source CFD/Adjoint optimisation framework¹

- Developed at Stanford University
- General purpose PDE solution methods
- Range of numerical schemes available (JST, ROE, MG, Euler-Implicit,...)
- Mesh deformation/adaptation
- Continuous Adjoint Solver

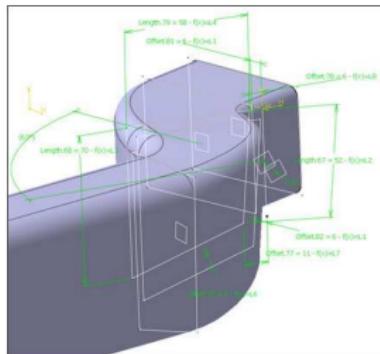


SU²
The Open-Source CFD Code

Native parameterisations in SU^2 

²images taken from <http://su2.stanford.edu/>

How to link any parameterisation to SU^2 ?



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1 Overview

2 Motivation

3 Gradient Calculation

4 Test Cases

5 Conclusions

Adjoint Based Optimisation

- Gradients required for optimisation
- Finite differences not feasible for complex shapes with multiple parameters
- Adjoint method provides an efficient alternative

Gradient Calculation

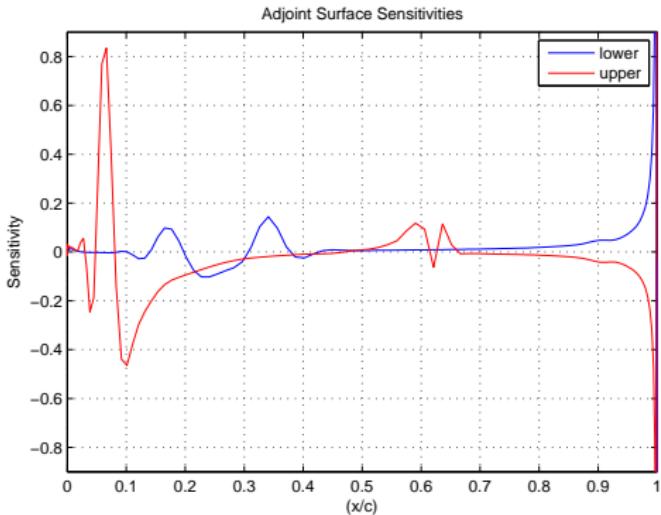
$$\left[\begin{array}{c} \frac{\partial f}{\partial A_1} \\ \frac{\partial f}{\partial A_2} \\ \vdots \\ \frac{\partial f}{\partial A_n} \end{array} \right] = \underbrace{\left[\begin{array}{ccc} \frac{\partial x_1}{\partial A_1} & \cdots & \frac{\partial x_m}{\partial A_1} \\ \vdots & \ddots & \vdots \\ \frac{\partial x_1}{\partial A_n} & \cdots & \frac{\partial x_m}{\partial A_m} \end{array} \right]}_{\text{Geometric Sensitivities}} \left[\begin{array}{c} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_m} \end{array} \right] \underbrace{\left[\begin{array}{c} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_m} \end{array} \right]}_{\text{Surface Sensitivities}}$$

- $\frac{\partial f}{\partial A_i}$ - Gradient
- $\frac{\partial x_j}{\partial A_i}$ - Geometric Sensitivities
- $\frac{\partial f}{\partial x_j}$ - Surface Sensitivities

Gradient Calculation

$$\left[\begin{array}{c} \frac{\partial f}{\partial A_1} \\ \frac{\partial f}{\partial A_2} \\ \vdots \\ \frac{\partial f}{\partial A_n} \end{array} \right] = \underbrace{\left[\begin{array}{ccc} \frac{\partial x_1}{\partial A_1} & \cdots & \frac{\partial x_m}{\partial A_1} \\ \vdots & \ddots & \vdots \\ \frac{\partial x_1}{\partial A_n} & \cdots & \frac{\partial x_m}{\partial A_m} \end{array} \right]}_{\text{Geometric Sensitivities}} \underbrace{\left[\begin{array}{c} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_m} \end{array} \right]}_{\text{Surface Sensitivities}}$$

- $\frac{\partial f}{\partial A_i}$ - Gradient
- $\frac{\partial x_j}{\partial A_i}$ - Geometric Sensitivities
- $\frac{\partial f}{\partial x_j}$ - Surface Sensitivities

SU^2 

Computed by SU^2

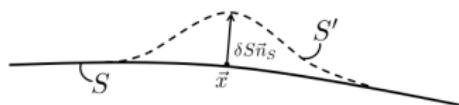
Gradient Calculation

$$\left[\begin{array}{c} \frac{\partial f}{\partial A_1} \\ \frac{\partial f}{\partial A_2} \\ \vdots \\ \frac{\partial f}{\partial A_n} \end{array} \right] = \underbrace{\left[\begin{array}{ccc} \frac{\partial x_1}{\partial A_1} & \cdots & \frac{\partial x_m}{\partial A_1} \\ \vdots & \ddots & \vdots \\ \frac{\partial x_1}{\partial A_n} & \cdots & \frac{\partial x_m}{\partial A_n} \end{array} \right]}_{\text{Geometric Sensitivities}} \left[\begin{array}{c} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_m} \end{array} \right] \underbrace{\left[\begin{array}{c} \frac{\partial f}{\partial x_1} \\ \frac{\partial f}{\partial x_2} \\ \vdots \\ \frac{\partial f}{\partial x_m} \end{array} \right]}_{\text{Surface Sensitivities}}$$

- $\frac{\partial f}{\partial A_i}$ - Gradient
- $\frac{\partial x_j}{\partial A_i}$ - Geometric Sensitivities
- $\frac{\partial f}{\partial x_j}$ - Surface Sensitivities

Gradient Calculation

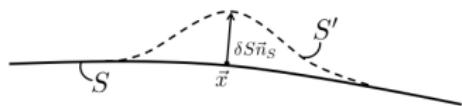
Use SU^2 native parameterisations



SU^2

Gradient Calculation

Use SU^2 native parameterisations



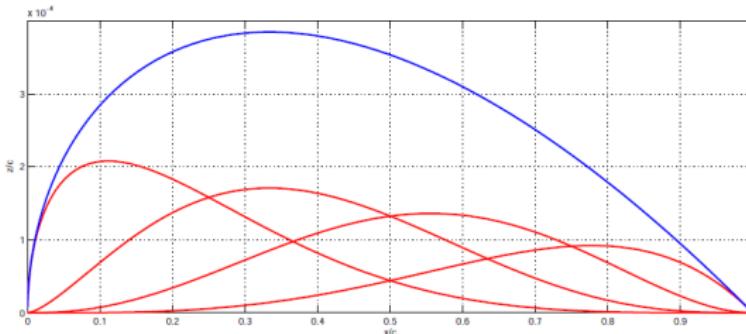
Gradient Calculation

Or add your own parametric sensitivities:

$$\Delta J = \int \phi V_n dS$$

CST Parameterisation

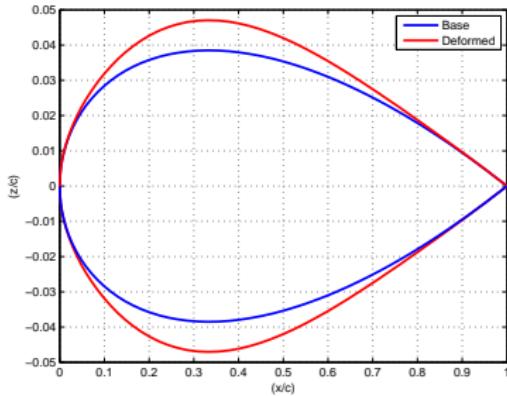
$$\begin{aligned}\zeta(\phi) &= C_{N2}^{N1}(\phi)S(\phi) + \phi\Delta\zeta_{TE} \\ C_{N2}^{N1}(\phi) &= \phi^{N1}(1-\phi)^{N2} && \text{Class Function} \\ S(\phi) &= \sum_{i=0}^n A_i S_i && \text{Shape Function}\end{aligned}$$



CST Parameterisation

The surface is manipulated through the choice of function weights A_i :

$$S(\phi) = \sum_{i=0}^n A_i S_i$$

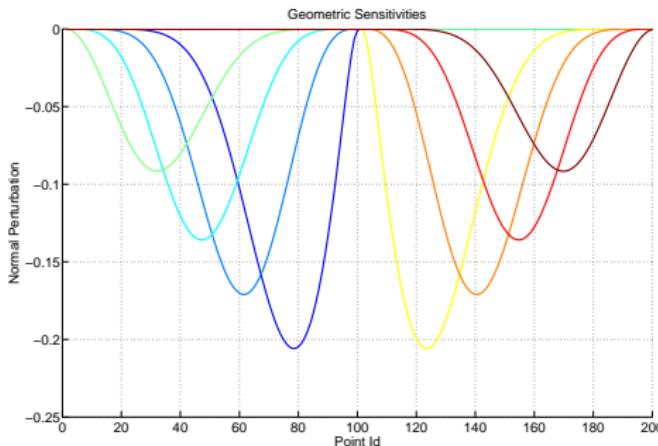


CST Parameterisation

Geometric Sensitivities ($\equiv V_n$)

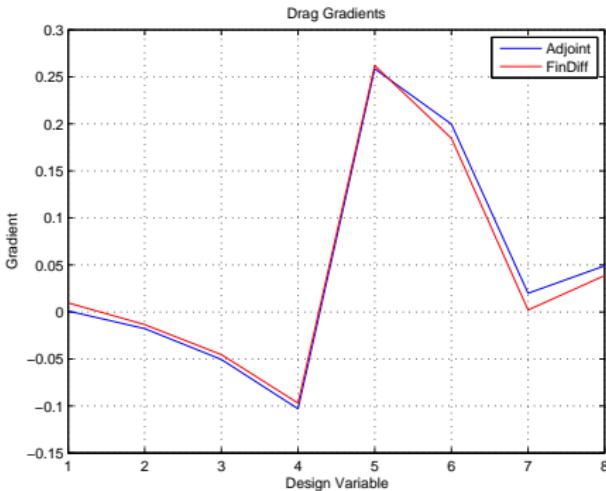
$$\frac{\partial \mathbf{x}_j}{\partial A_i} - \text{Geometric Sensitivities}$$

$$\frac{\partial \mathbf{x}_j}{\partial A_i} = \left(\frac{\partial x_j}{\partial A_i} n_x + \frac{\partial y_j}{\partial A_i} n_y + \frac{\partial z_j}{\partial A_i} n_z \right)$$

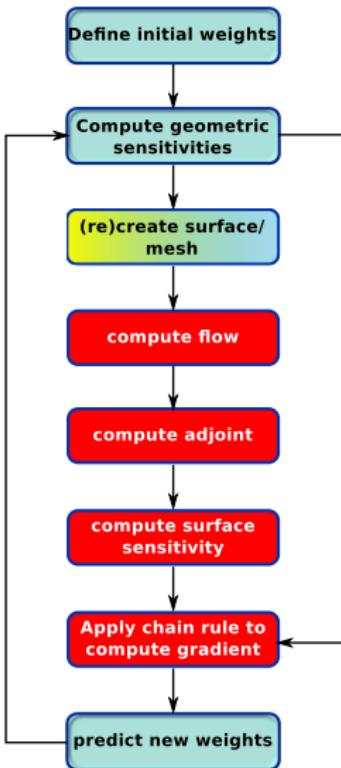


Validation - CST parameterisation

$\frac{\partial f}{\partial A_i}$ - Gradient



SU^2 Optimisation Process



Outline

1 Overview

2 Motivation

3 Gradient Calculation

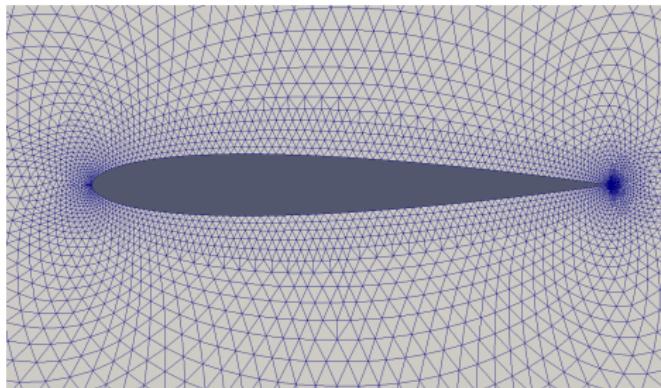
4 Test Cases

5 Conclusions

NACA0012 Drag minimization

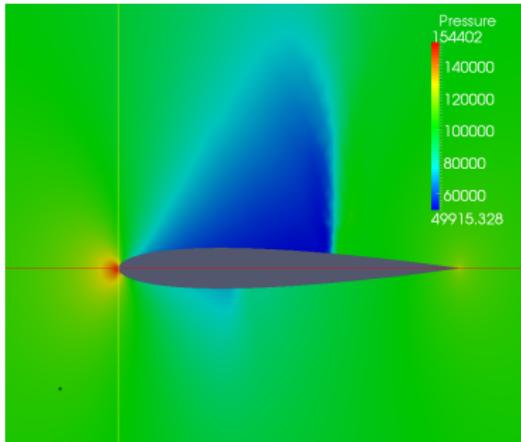
Inviscid test case: NACA0012 starting aerofoil.

- $M_\infty = 0.8$
- $\alpha = 1.25^\circ$
- $f = \min(C_d)$
- $C_l \geq 0.33$
- $C_m > 0.034$
- nDV = 8



NACA0012 Drag minimization

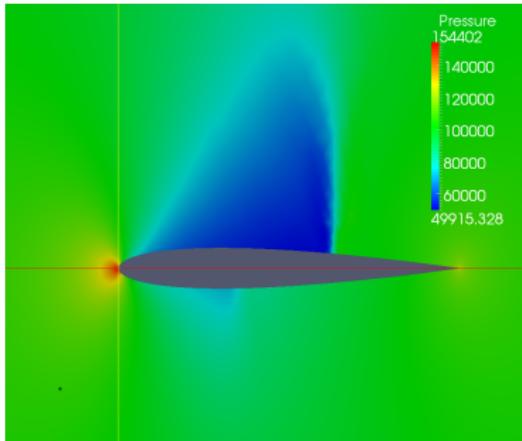
Inviscid test case: NACA0012 starting aerofoil.



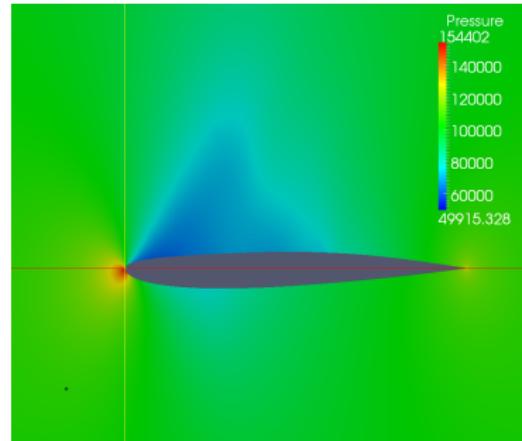
initial aerofoil

NACA0012 Drag minimization

Inviscid test case: NACA0012 starting aerofoil.



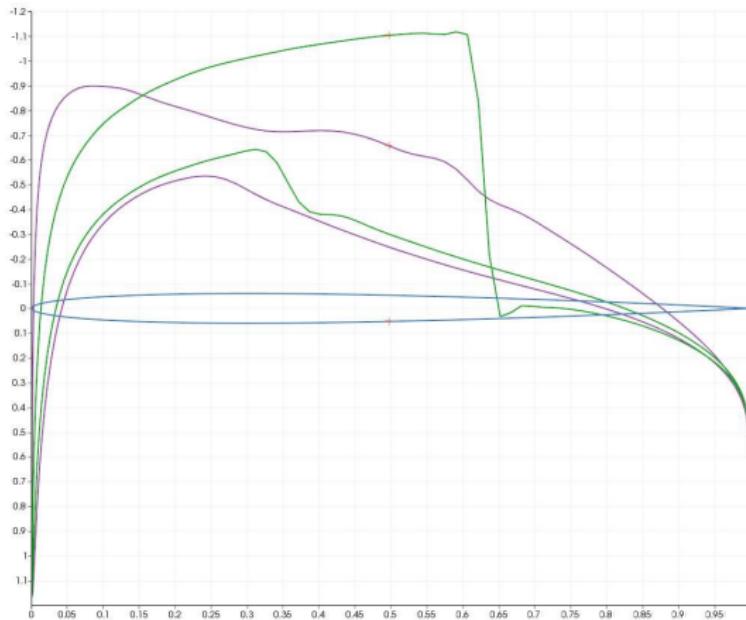
initial aerofoil



final aerofoil

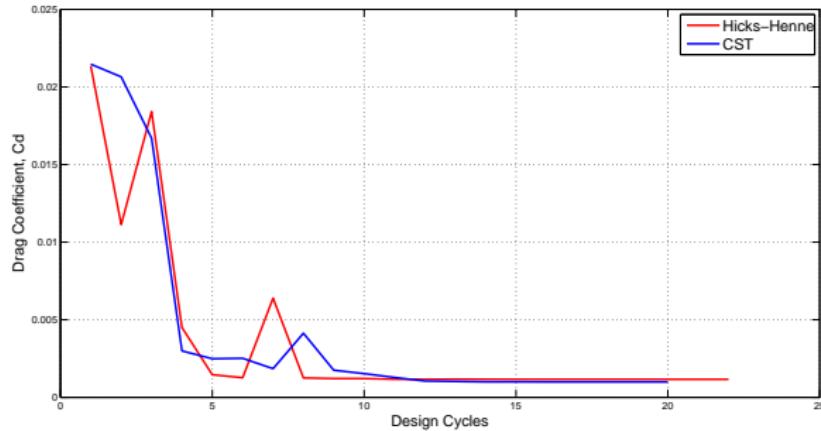
NACA0012 Drag minimization

Inviscid test case: NACA0012 starting aerofoil.



NACA0012 Drag minimization

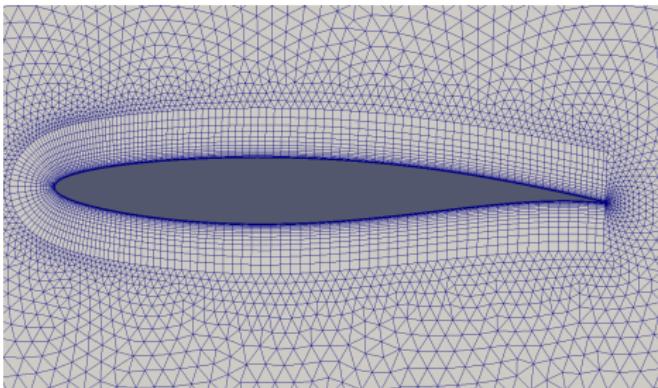
Inviscid test case: NACA0012 starting aerofoil.



RAE2822 Drag minimization

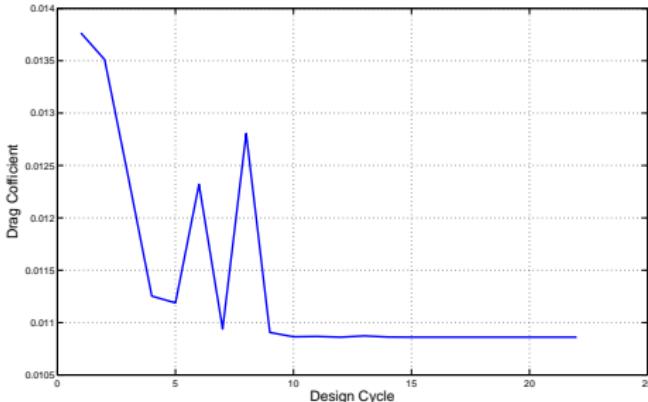
Viscous test case: RAE2822 starting aerofoil.

- $M_\infty = 0.729$
- $\alpha = 2.31^\circ$
- $f = \min(C_d)$
- $nDV = 8$
- SA turbulence model
- $y^+ \leq 5$



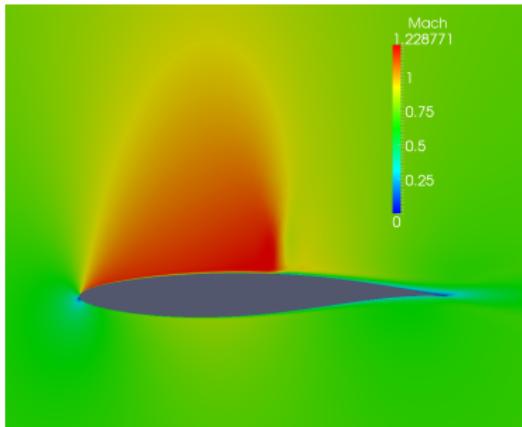
RAE2822 Drag minimization

Viscous test case: RAE2822 starting aerofoil.



RAE2822 Drag minimization

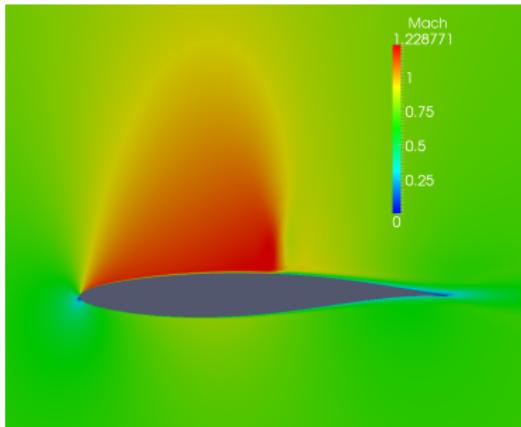
Viscous test case: RAE2822 starting aerofoil.



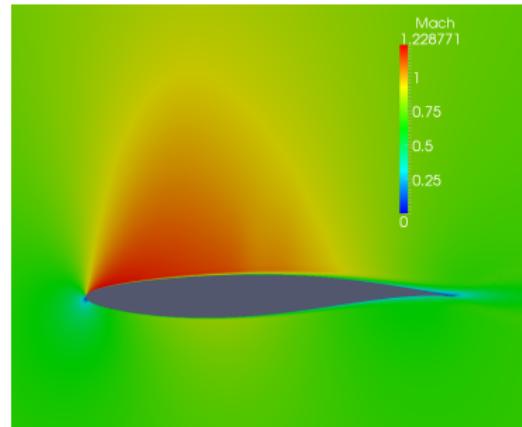
initial aerofoil

RAE2822 Drag minimization

Viscous test case: RAE2822 starting aerofoil.



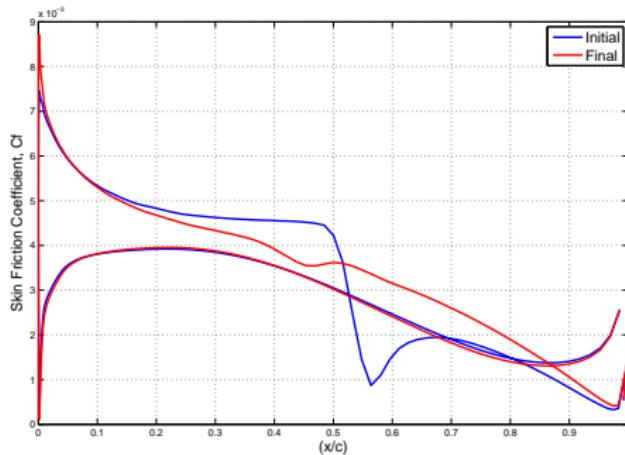
initial aerofoil



final aerofoil

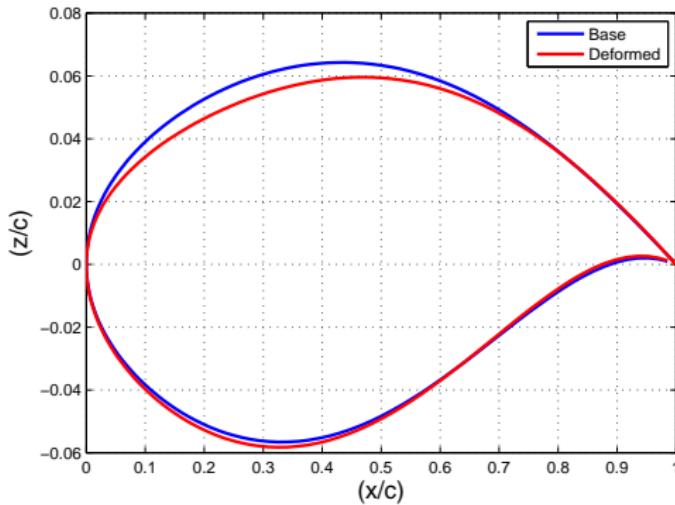
RAE2822 Drag minimization

Viscous test case: RAE2822 starting aerofoil.



RAE2822 Drag minimization

Viscous test case: RAE2822 starting aerofoil.



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RAE2822 Drag minimization

- An alternative parameterisation was introduced into the SU^2 suite
- Model deformation can be performed outside SU^2
- alternative approach does not compromise optimisation efficiency with respect to native parameterisations

Q & A

Thank you for your attention

Questions Welcome

CST Parameterisation

$$\begin{aligned}\zeta(\phi) &= C_{N2}^{N1}(\phi)S(\phi) + \phi\Delta\zeta_{TE} \\ C_{N2}^{N1}(\phi) &= \phi^{N1}(1 - \phi)^{N2} \\ S(\phi) &= \sum_{i=0}^n A_i S_i \\ S_i &= K_{i,n} \phi^i (1 - \phi)^i \\ K_{i,n} &= \binom{n}{i} = \frac{n!}{i!(n - i)!}\end{aligned}$$