Profile Optimization Method for Robust Airfoil Shape Optimization in Viscous Flow

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Abstract

Simulation results obtained by using FUN2D for robust airfoil shape optimization in transonic viscous flow are included to show the potential of the profile optimization method for generating fairly smooth optimal airfoils with no off-design performance degradation.

Nomenclature

- $c$: chord length of airfoil
- $c_d$: drag coefficient
- $\frac{\partial c_d}{\partial D}$: gradient of $c_d$ with respect to $D$
- $\frac{\partial c_d}{\partial \alpha}$: derivative of $c_d$ with respect to $\alpha$
- $c_l$: lift coefficient
- $\frac{\partial c_l}{\partial D}$: gradient of $c_l$ with respect to $D$
- $\frac{\partial c_l}{\partial \alpha}$: derivative of $c_l$ with respect to $\alpha$
- $c_{l*}$: target lift coefficient
- $D$: design vector
- $E(\cdot)$: mean of random variable
- $\mathcal{F}$: feasible set for the design vector $D$
- $M$: free-stream Mach number
- $n$: number of design variables
- $p(M)$: probability density function of Mach number
- $r$: number of design conditions
- $x, y$: coordinates of points on plane
- $\alpha$: angle of attack
- $\gamma_{\text{min}}$: minimum rate of drag reduction at all design conditions
- $\Delta D$: change in design vector
- $\delta_{k, i}, \rho_k$: scalars defining the trust region
- $\sigma^2(\cdot)$: variance of random variable
- $\Omega$: a given Mach range
- $\langle \cdot, \cdot \rangle$: inner product in Euclidean space

Subscripts and Superscripts

- $i$: index for design condition
- $j$: index for component of design vector
- $k$: index for iteration or iterate

1 Introduction

This paper is an extension of reference 1, where the profile optimization method was proposed to resolve off-design performance degradation problems encountered by multipoint
optimization methods (ref. 2). Numerical simulation results given in reference 1, though very encouraging, were not representative of realistic airfoil shape optimization problems. To strengthen the simulation results in reference 1, we apply the profile optimization method to two cases of airfoil shape optimization in transonic viscous flow over a range of Mach numbers.

The first case is the redesign of the RAE2822 airfoil (ref. 2) over the Mach range from 0.68 to 0.76, with the target lift at 0.733 and Reynolds number $2.7 \times 10^6$. The second case is the redesign of Whitcomb's integral supercritical airfoil (ref. 3) over the Mach range from 0.68 to 0.77, with the target lift at 0.7 and Reynolds number $2.7 \times 10^6$. Viscous flow analysis and the corresponding gradient calculation are based on FUN2D (ref. 4), which can perform fully turbulent Navier-Stokes flow analysis and the corresponding discrete adjoint analysis with unstructured grids. Two spar location thickness constraints and the maximum thickness constraint are enforced for both airfoil shape optimization cases. Airfoils are parameterized by 35 B-spline control points. Design variables are angles of attack at specified design conditions and the y-coordinates of 35 B-spline control points.

In both cases, the simulation results show that the profile optimization method can generate fairly realistic optimal airfoil shapes in transonic viscous flow without severe off-design performance degradation, when 4 design conditions and 35 geometric design variables are used.

Because the purpose of this paper is to provide additional simulation results for the profile optimization method given in reference 1, readers should consult reference 1 for relevant references and technical details.

The paper is organized as follows. In section 2, we review the profile optimization method. Section 3 includes the simulation results for airfoil shape optimization in transonic viscous flow. Concluding remarks are given in the final section.

2 Profile Optimization Method

In this section, we briefly review the profile optimization method proposed in reference 1. The profile optimization method is intended to solve the following robust optimization problem:

$$
\min_{D, \alpha(M)} \left( E(c_d), \sigma(c_d) \right)
$$

subject to

$$
D \in \mathcal{F} \quad \text{and} \quad c_l(D, \alpha(M), M) = c_l^*(M) \quad \text{for} \ \ M \in \Omega.
$$

Here $c_l^*(M)$ is the target lift requirement for Mach number $M$, $\mathcal{F}$ is a given feasible set for geometric design variables (that could be defined by geometry constraints such as thickness constraints), and $\alpha(M)$ is the angle of attack corresponding to $M$. The drag and lift coefficients are $c_d$ and $c_l$, respectively. The mean and variance of $c_d$ are defined as

$$
E(c_d) = \int_{\Omega} c_d(D, \alpha(M), M) \cdot p(M) \ dM,
$$

$$
\sigma^2(c_d) = \int_{\Omega} \left[ c_d(D, \alpha(M), M) - E(c_d) \right]^2 p(M) \ dM,
$$

where $p(M)$ is a probability density function of $M$ and $\Omega$ is a given Mach range (such as from $M = 0.68$ to $M = 0.77$).

The robust optimization model in equation (1) addresses some important issues in aerodynamic shape optimization. For example, to avoid off-design performance degradation, one can reduce $\sigma(c_d)$ as much as possible. Note that if $\sigma(c_d) = 0$, the corresponding solution will

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have the same (perhaps poor) performance over the given Mach range. Therefore, equation (1) provides a tool for trade-offs between average performance improvement and performance fluctuations over the Mach range. However, due to the high computational cost for solving Navier-Stokes equations, it is impossible to have reasonable estimates of \( E(c_d) \) and \( \sigma(c_d) \). Consequently, it is not practical to solve equation (1) using current computational fluid dynamics tools.

One alternative way to avoid off-design performance degradation is to find a descent direction that could reduce the drag \textit{simultaneously and proportionally} over the given Mach range while keeping the lift at the target value. The profile optimization method was designed to find such a descent direction with limited information on lift and drag over the given Mach range. The innovative feature of the profile optimization method is to adaptively change the objective function from iteration to iteration to achieve simultaneous and proportional drag reduction over the given Mach range. In contrast to methods that minimize one aggregate objective function to find a Pareto optimal solution to a multiobjective optimization problem, the profile optimization method does not solve any optimization problem with one objective function; instead, it looks for a Pareto optimal solution that has the least chance for severe off-design performance degradation.

**Profile Optimization Method.** Let \( D^0 \) be a given initial design vector, let \( M_1, \ldots, M_T \) be a set of design points over the given Mach range, and \( k = 0 \). Construct a sequence of design vectors as follows:

1. Compute feasible angles of attack. Find \( \alpha_{1,k}, \alpha_{2,k}, \ldots, \alpha_{r,k} \) such that
   \[
   c_d(D^k, \alpha_{i,k}, M_i) = c^*_d \quad \text{for } 1 \leq i \leq r.
   \]

2. Solve a trust region subproblem. Let \( c_{d,i,k} \) and \( c_{l,i,k} \) be the linear approximations of the drag and lift at \( (D^k, \alpha_{i,k}) \):
   \[
   c_{l,i,k}(\Delta D, \Delta \alpha_i) = c_l(D^k, \alpha_{i,k}, M_i) + \left( \frac{\partial c_l}{\partial D}, \Delta D \right) + \frac{\partial c_l}{\partial \alpha_i} \Delta \alpha_i,
   \]
   \[
   c_{d,i,k}(\Delta D, \Delta \alpha_i) = c_d(D^k, \alpha_{i,k}, M_i) + \left( \frac{\partial c_d}{\partial D}, \Delta D \right) + \frac{\partial c_d}{\partial \alpha_i} \Delta \alpha_i,
   \]
   where the derivatives are evaluated at \( (D^k, \alpha_{i,k}, M_i) \).

   Consider the following trust region subproblem:
   \[
   \min_{\Delta D, \Delta \alpha_i} \quad \eta \quad \text{subject to} \quad D^k + \Delta D \in \mathcal{F}, \quad (4)
   \]
   \[
   -\delta_{i,k}\rho_k \leq \Delta D_i \leq \delta_{i,k}\rho_k \quad \text{for } 1 \leq i \leq n,
   -\alpha_k \leq \Delta \alpha_i \leq \alpha_k \quad \text{for } 1 \leq i \leq r,
   c_{l,i,k}(\Delta D, \Delta \alpha_i) = c^*_l \quad \text{for } 1 \leq i \leq r,
   c_{d,i,k}(\Delta D, \Delta \alpha_i) \leq (1 - \eta) \cdot c_d(D^k, \alpha_{i,k}, M_i) \quad \text{for } 1 \leq i \leq r,
   \]
   where \( \delta_{i,k} \geq 0 \) and \( \alpha_k > 0 \) are scalars that determine the trust region, and \( D^k + \Delta D \in \mathcal{F} \) means that the airfoil corresponding to \( (D^k + \Delta D) \) satisfies all the geometric constraints. (For our simulation runs, \( \delta_{i,k} \) is approximately the thickness of the airfoil at the \( x \)-coordinate of the \( i \)-th control point, \( \alpha_k = 1 \), and \( D^k + \Delta D \in \mathcal{F} \) means that the airfoil satisfies two thickness constraints at the spar locations and the maximum thickness constraint.) Determine the smallest \( \rho_k > 0 \) such that equation (4) is feasible and the least norm solution \( (\Delta D^k, \Delta \alpha_{1,k}, \ldots, \Delta \alpha_{r,k}) \) of equation (4) satisfies the following condition:
   \[
   c_{d,i,k}(\Delta D^k, \Delta \alpha_{i,k}) \leq (1 - \gamma_{\min})c_d(D^k, \alpha_{i,k}, M_i) \quad \text{for } 1 \leq i \leq r.
   \]
3. Generate the new iterate. Let \( \alpha_{i,k+1} = \alpha_{i,k} + \Delta \alpha_{i,k} \) for \( 1 \leq i \leq r \) and \( D^{k+1} = D^k + \Delta D^k \).

4. Start a new iteration. Update \( k \) by \( k + 1 \) and go back to step 1.

3 Numerical Simulation Results

We apply the profile optimization method in two cases of airfoil shape design optimization in transonic viscous flow. The first case is the redesign of the RAE2822 airfoil over the Mach range from 0.68 to 0.76. We adopt the four design conditions used in Drela's simulation (ref. 2): \( M = 0.68, 0.71, 0.74, \) and \( 0.76 \), with the target lift at 0.733 and Reynolds number \( 2.7 \times 10^6 \). The second case is the redesign of Whitcomb's integral supercritical airfoil (ref. 3) over the Mach range from 0.68 to 0.77. In this case, we use the following four design conditions: \( M = 0.68, 0.71, 0.74, \) and \( 0.77 \), with the target lift at 0.7 and Reynolds number \( 2.7 \times 10^6 \).

For a given airfoil, the lift and drag coefficients and their gradients are calculated by solving fully turbulent Navier-Stokes equations and the corresponding discrete adjoint equations using FUN2D (ref. 4). In both cases, the unstructured grid has 300 grid points on the airfoil and 32 grid points on the far field (which is placed at 20 chord lengths). The unstructured grid has a total of 18654 grid points, 55631 elements, and 37308 faces (see fig. 1 for the grid near the RAE2822 airfoil).

With the given grid, the flow solver and the adjoint solver are terminated when the 2-norms of the residual of the density equations and the corresponding discrete adjoint equations are reduced by at least five orders of magnitude.

Airfoils are parameterized by 35 B-spline control points, which is large enough to accommodate free-form geometry changes for both cases. Figure 2 shows parameterization of the RAE2822 airfoil and Whitcomb’s integral supercritical airfoil.

The \( x \)-coordinates of all the control points are fixed during optimization. Changes of the \( y \)-coordinates of the five control points near the trailing edge are constrained to be the same so that the shape of the airfoil near the trailing edge is not too oscillatory. The same geometry constraint is applied to the \( y \)-coordinates of the three control points near the leading edge. The reason for such geometric constraints is that the optimizer may not be able to make a reasonable modification of the shape of the leading or trailing edge at very fine scales. All the \( y \)-coordinates of the 35 control points are used as shape design variables. In both cases, we impose a thickness constraint at spar locations \( x/c = 0.15 \) and \( x/c = 0.6 \) and at the maximum thickness location. For each iterate \( D^k \), we find angles of attack \( \alpha_{1,k}, \ldots, \alpha_{r,k} \) such that \( |c_l(D^k, \alpha_{i,k}, M_i) - c_l|/c_l \leq 0.001 \).

3.1 Drag Reduction Over the Given Mach Range

Figures 3 and 4 show the history of drag changes at the design conditions for cases 1 and 2, respectively.

Figure 5 shows the postoptimization analysis of drag curve over the given Mach range for cases 1 and 2. The drag rise curves are constructed by using 10 equally spaced Mach numbers from \( M = 0.68 \) to \( M = 0.77 \), which include 4 design points and 6 off-design points. In both cases, the profile optimization method reduces the drag of the baseline over the given Mach range. Moreover, there is no off-design performance degradation of the optimized airfoils.
3.2 Airfoil Shapes

The optimal airfoils generated by the profile optimization method are quite smooth and do not have shock bumps corresponding to design conditions (i.e., Mach numbers). This result is different from the numerical simulation results obtained earlier by Drela (ref. 2) on multipoint airfoil optimization when sinusoidal basis functions were used to parameterize the airfoil shape. Figure 6 shows the optimal airfoils generated by the profile optimization method for cases 1 and 2.

3.3 Pressure Distributions

Pressure distributions for the baseline and the optimal airfoil in both cases are plotted in figures 7 and 8.

4 Concluding Remarks

This paper documents two cases of using the profile optimization method to redesign airfoils over a range of Mach numbers in transonic viscous flow with airfoils parameterized by 35 B-spline control points. FUN2D (ref. 4) is used to compute the lift and drag values and their gradients with respect to changes in airfoil shape and changes in angle of attack. All the y-coordinates of 35 B-spline control points are used by the profile optimization method to search for the true optimal solution. With thickness constraints at two spar locations and the maximum thickness constraint, the profile optimization method minimizes the drag at four design conditions while keeping the lift at the target value. The optimized airfoils have fairly realistic and smooth shapes. Postoptimization analysis shows that the optimized airfoils have no off-design performance degradation.

The simulation results demonstrate that the profile optimization method has the potential to be used for real-world airfoil shape optimization over a range of flight conditions.

References


Figure 1: Grid used to solve Navier-Stokes equations by FUN2D.

Figure 2: RAE2822 airfoil and Whitcomb's integral supercritical airfoil parameterized by 35 cubic B-spline control points.
Figure 3: Changes of drag counts at four design conditions for case 1 with target lift 0.733.

Figure 4: Changes of drag counts at four design conditions for case 2 with target lift 0.7.
Figure 5: Postoptimization analysis of drag over given Mach range for case 1 with target lift 0.733 and case 2 with target lift 0.7.

Figure 6: Optimal airfoils generated by profile optimization method.
Figure 7: Pressure distributions on RAE2822 and the corresponding optimal airfoil at four design conditions with target lift at 0.733.

Figure 8: Pressure distributions on Whitcomb’s integral supercritical airfoil and the corresponding optimal airfoil at four design conditions with target lift at 0.7.
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