

Unsteady Discrete Adjoint Formulation for High-order Discontinuous Galerkin Discretizations in Time-dependent Flow Problems

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This paper presents an unsteady discrete adjoint algorithm for high-order implicit discontinuous Galerkin discretizations in time-dependent flow problems. The major function of the adjoint approach is to obtain the sensitivity information in a time-dependent functional output, which in turn is used to drive an unsteady shape optimization process to deliver a minimum of the objective functional. A gradient-based optimization strategy is investigated, in which the sensitivity derivatives of the objective functional with respect to input variables are formulated in the context of high-order discontinuous Galerkin discretizations while special emphasis is given to the variations and linearizations of curvilinear boundary elements. Implicit temporal discretizations consisting of a second-order backward Euler (BDF2) scheme and a fourth-order implicit Runge-Kutta (IRK4) scheme are considered exclusively in this work, where the corresponding adjoint problem is required to be solved in a backward time-integration manner due to the associated transpose operation. Two numerical examples for the unsteady shape design techniques are presented to verify the derived sensitivity formulations and to demonstrate the performance of the adjoint approach, where the first involves an inverse shape optimization case by matching a time-dependent target pressure profile for a two-dimensional periodic vortical gust impinging on a RAE-2822 airfoil, and the second considers minimization of the acoustic noise produced by subsonic flow over a NACA0012 airfoil with a $0.03c$ blunt trailing edge.

I. Introduction

High-order discontinuous Galerkin (DG) methods have become a popular approach over the last decade in solving a variety of computational fluid dynamics problems¹⁻⁶. The advantage of high-order DG discretizations is that they alter the asymptotic relationship between solution accuracy and resolution in a beneficial manner. Over the same period, the incorporation of sensitivity analysis techniques⁷⁻¹³ has become the subject of increased interest in many numerical simulations. The sensitivities of specific simulation outputs with respect to simulation inputs deliver important information in a simulation which in turn can be used to drive a design optimization process^{14,15} to obtain a minimum of an objective functional. In addition, adjoint methods in gradient-based design optimization techniques enable efficient calculation of the sensitivity of an objective with respect to any number of design variables at a cost which is essentially independent of the number of design variables, and roughly equivalent to an additional flow analysis problem^{16,17}. While the majority of work in adjoint-based aerodynamic shape optimizations has been focused on the use of finite-volume methods in a steady flow environment¹⁷⁻²⁰, relatively little work has been done with regards to the use of high-order DG methods for shape designs in an unsteady flow environment^{21,22}. Therefore, the goal of this work is to develop an unsteady discrete adjoint algorithm for high-order discontinuous Galerkin discretizations in time-dependent flow problems and to apply this technique to unsteady shape optimization problems.

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It is known that high-order methods require the use of curved boundary elements to deliver an overall high-accuracy solution^{3,23,24}. This is especially the case for problems with complicated geometries where high-order methods typically use relatively coarse or large boundary elements compared to lower-order methods. Therefore this work uses additional surface quadrature points^{3,12} on physical (wall) boundaries to determine the geometry mappings of high-order accurate geometric surface shapes. In a shape optimization process of this circumstance, when design variables produce changes in the shape of the domain boundaries, the additional face quadrature points must be deformed in the same manner as the surface nodes. Accordingly, the linearizations of the discretized unsteady flow equations as well as the objective functionals with respect to the mesh geometry must consider the variation from both mesh grid points and extra surface quadrature points in the element mappings. The present work focuses on the development of the techniques required to incorporate these effects and results are demonstrated in two unsteady shape optimization problems.

In each typical optimization cycle, numerical approximations for the time-dependent flow problem and adjoint problem are required, where the unsteady flow solution is solved in a forward integration in time, however, due to the transpose operator in the discrete adjoint formulations, the unsteady adjoint solution which corresponds to a series of linear problems spanning the entire time domain is solved after the flow analysis problem, but in a backward time-integration manner. To avoid the restriction of time-step sizes of explicit schemes^{25,26}, we make exclusive use of implicit temporal schemes, consisting of a second-order backwards difference Euler (BDF2) scheme and a fourth-order implicit Runge-Kutta (IRK4) scheme²⁷. It is noted that the difference of the adjoint solution procedure between the IRK4 scheme and the BDF2 scheme lies in the fact that the adjoint variables in the IRK4 scheme are required to be computed at both the discrete time step locations and all the intermediate stages for each implicit time step.

Additionally, one of the flow problems of interest in the present work consists of a single airfoil gust response case taken from the computational aeroacoustics (CAA) benchmark problems²⁸, in which a highly-accurate time-dependent solution is required to capture the airfoil unsteady pressure, the noise generation and radiation produced by a periodic vortical gust impinging on an airfoil. This type of the gust interaction problem has many practical applications such as turbomachines or helicopter rotors, and hence a series of computational aeroacoustics solvers²⁹⁻³¹ have been developed in the past. This work investigates current capabilities of the high-order implicit DG solver for the aeroacoustics problem and applies the aeroacoustics problem to an unsteady airfoil shape optimization case.

The outline of this paper is as follows. In Section II the governing equations are introduced and the spatial discontinuous Galerkin discretizations as well as implicit time-integration schemes are formulated. Section III focuses on the formulation of discrete adjoint-based sensitivity derivatives in the context of high-order discontinuous Galerkin methods for unsteady shape optimization problems, with special emphasis given on the variation and re-creation of high-order curved boundary or surface elements. The validation of the present DG solver for the CAA benchmark problem is first provided in Section IV, followed by two numerical examples for verifying the unsteady discrete adjoint formulations derived in this work and for demonstrating the performance of the unsteady adjoint techniques. Finally, Section V summarizes the conclusions of the current work and discusses some possible directions for future work.

II. Governing Equations and Discretizations

The governing equations that we consider exclusively in this work are the two-dimensional compressible Euler equations that can be written in the following conservative form:

$$\frac{\partial \mathbf{u}(\mathbf{x}, t)}{\partial t} + \frac{\partial \mathbf{f}_1(\mathbf{u}(\mathbf{x}, t))}{\partial x} + \frac{\partial \mathbf{f}_2(\mathbf{u}(\mathbf{x}, t))}{\partial y} = 0 \quad \text{in } \Omega \quad (1)$$

where Ω is a two-dimensional bounded domain. The vector of conservative flow variables \mathbf{u} and the inviscid Cartesian flux components \mathbf{f}_1 and \mathbf{f}_2 are defined by

$$\mathbf{u} = \begin{Bmatrix} \rho \\ \rho u \\ \rho v \\ \rho e \end{Bmatrix}, \quad \mathbf{f}_1 = \begin{Bmatrix} \rho u \\ \rho u^2 + p \\ \rho uv \\ (\rho e + p)u \end{Bmatrix}, \quad \mathbf{f}_2 = \begin{Bmatrix} \rho v \\ \rho uv \\ \rho v^2 + p \\ (\rho e + p)v \end{Bmatrix} \quad (2)$$

respectively, where the notations ρ , p , and e denote the fluid density, pressure and specific total energy per unit mass, respectively. u and v represent the flow velocity components in the x and y coordinate directions. This system of equations is completed by the perfect gas equation of state given as,

$$p = (\gamma - 1) \left[\rho e - \frac{1}{2} \rho (u^2 + v^2) \right] \quad (3)$$

where γ is defined as the ratio of specific heats, which is 1.4 for air.

A. Discontinuous Galerkin Discretizations

The computational domain Ω is partitioned into an ensemble of non-overlapping elements, such that $\Omega = \bigcup_k \Omega_k$, where Ω_k refers to the volume of an element k in the computational mesh. The discontinuous Galerkin discretization proceeds by formulating a weak statement of the governing equations, by multiplying Eq. (1) by a set of test functions, $\{\phi_j, j = 1, \dots, M\}$, with the maximum polynomial order of p , and integrating within each element, e.g. k , as:

$$\int_{\Omega_k} \phi_j \left[\frac{\partial \mathbf{u}}{\partial t} + \frac{\partial \mathbf{f}_1(\mathbf{u})}{\partial x} + \frac{\partial \mathbf{f}_2(\mathbf{u})}{\partial y} \right] d\Omega_k = 0 \quad (4)$$

Integrating this equation by parts, the weak statement of the problem becomes:

$$\begin{aligned} & \int_{\Omega_k} \phi_j \frac{\partial \mathbf{u}_p}{\partial t} d\Omega_k - \int_{\Omega_k} \left[\frac{\partial \phi_j}{\partial x} \mathbf{f}_1(\mathbf{u}_p) + \frac{\partial \phi_j}{\partial y} \mathbf{f}_2(\mathbf{u}_p) \right] d\Omega_k \\ & + \int_{\partial\Omega_k \setminus \partial\Omega} \phi_j \mathbf{H}(\mathbf{u}_p^+, \mathbf{u}_p^-, \mathbf{n}) dS + \int_{\partial\Omega_k \cap \partial\Omega} \phi_j \mathbf{H}^b(\mathbf{u}_p^-, \mathbf{u}_p^+, \mathbf{n}) dS = 0 \end{aligned} \quad (5)$$

where the unit normal vector $\mathbf{n} = (n_x, n_y)$ is outward to the boundary. \mathbf{u}_p^+ and \mathbf{u}_p^- refer to the interior and exterior Galerkin solution approximations at shared inter-element interfaces, respectively. Current implementations of the interior boundary flux function $\mathbf{H}(\mathbf{u}_p^+, \mathbf{u}_p^-, \mathbf{n})$ include the Riemann flux approximation of HLLC³² and Lax-Friedrichs³³. For edges coinciding with the computational domain boundaries, the approximate flux function, $\mathbf{H}^b(\mathbf{u}_p^-, \mathbf{u}_p^+, \mathbf{n})$, is required to be explicitly dependent on exterior traces to satisfy the dual-consistency condition³⁴. Here we set $\mathbf{H}^b(\mathbf{u}_p^-, \mathbf{u}_p^+, \mathbf{n}) = \mathbf{f}_1(\mathbf{u}_p^-)n_x + \mathbf{f}_2(\mathbf{u}_p^-)n_y$, where \mathbf{u}_p^- is determined by the interior flow approximation, \mathbf{u}_p^+ , as well as the given boundary conditions. In particular, total temperature and pressure are prescribed on the inflow boundary and static pressure is prescribed on the outflow boundary, and at solid walls \mathbf{u}_p^- is set to have the same density, total energy and tangential velocity $(u, v)_{\parallel}^t$ as \mathbf{u}_p^+ , given by:

$$(u, v)_{\parallel}^t = (u^+, v^+)^t - (u^+n_x + v^+n_y)(n_x, n_y)^t \quad (6)$$

The Galerkin finite-element approximation in the weak formulation is expanded as a series of truncated basis functions²⁷ and solution coefficients as:

$$\mathbf{u}_p = \sum_{i=1}^M \tilde{\mathbf{u}}_{p_i} \phi_i(\mathbf{x}) \quad (7)$$

Since the set of basis functions is defined in a reference triangle $\bar{\Omega}$ spanning between $\{0 < \xi, \eta < 1\}$ ³⁵, a coordinate mapping from the reference to a physical triangle, illustrated by Fig. 1, is required for the computation of the first-order derivatives and integrals denoted in Eq. (5). The reference-to-physical transformation and the corresponding Jacobian J_k associated with each element k are given by:

$$\mathbf{x}_k = \sum_{i=1}^M \tilde{\mathbf{x}}_{k_i} \phi_i(\xi, \eta) \quad J_k = \begin{bmatrix} \frac{\partial x}{\partial \xi} & \frac{\partial x}{\partial \eta} \\ \frac{\partial y}{\partial \xi} & \frac{\partial y}{\partial \eta} \end{bmatrix} \quad (8)$$

In the simple case of straight-sided elements the transformation is linear thus the geometric mapping $\tilde{\mathbf{x}}_k$ can be evaluated only by using the element vertex coordinates. However, in the more complex cases of high-order curved elements, which are typically required at curved wall boundaries even for inviscid flows, additional surface nodes³⁶ must be included for determining the higher-order modes ($p > 1$) of the geometric mapping coefficients, obtained by:

$$\tilde{\mathbf{x}}_k = \Phi^{-1} \hat{\mathbf{x}}_{p_k} \quad (9)$$

where $\hat{\mathbf{x}}_{p_k} = \{\mathbf{x}_{c_k}, \mathbf{x}_{q_k}\}$ refers to the coordinates of physical quadrature points, consisting of the element vertices \mathbf{x}_{c_k} as well as additional surface points \mathbf{x}_{q_k} for the element k . Φ denotes the projection mapping matrix which is constituted

by the basis functions evaluated at the aforementioned quadrature points ($\hat{\xi}_{pk} \leftarrow \hat{\mathbf{x}}_{pk}$) in the reference triangle. The evaluation of the volume integrals in Eq. (5) is computed by use of Gaussian quadrature rules³⁷ which are exact for polynomial degree $2p$, and the surface integrals use Gauss-Legendre-Lobatto quadrature rules³⁵ which are exact for polynomial degree of $2p + 1$ ^{3,27}.

Eq. (5) can be written in the following ordinary differential equation (ODE) form as:

$$M \frac{d\tilde{\mathbf{u}}_p}{dt} + \mathbf{R}(\tilde{\mathbf{u}}_p) = 0 \quad (10)$$

where \mathbf{R} represents discretized spatial residual and M denotes the mass matrix²⁷. Next we proceed to integrate in time using implicit temporal schemes.

B. Temporal Discretizations

To avoid the restriction of time-step sizes of explicit schemes, we make exclusive use of implicit temporal schemes, consisting of a second-order accurate backwards difference Euler scheme (BDF2) and a fourth-order implicit Runge-Kutta scheme (IRK4)^{27,38}. The formulations for the BDF2 and IRK4 schemes can be derived by starting from the set of ODEs given by Eq. (10), written as:

$$\text{BDF2: } \mathbf{R}_e^{n+1}(\tilde{\mathbf{u}}_p^{n+1}) = \frac{M}{\Delta t} \left(\frac{3}{2} \tilde{\mathbf{u}}_p^{n+1} \right) + \mathbf{R}(\tilde{\mathbf{u}}_p^{n+1}) - \frac{M}{\Delta t} \left(2\tilde{\mathbf{u}}_p^n - \frac{1}{2} \tilde{\mathbf{u}}_p^{n-1} \right) = 0 \quad (11)$$

$$\text{IRK4: } \mathbf{R}_e^{n+1}(\tilde{\mathbf{u}}_p^{(s),n+1}) = \frac{M}{\Delta t} \tilde{\mathbf{u}}_p^{(s),n+1} + a_{ss} \mathbf{R}(\tilde{\mathbf{u}}_p^{(s),n+1}) - \left[\frac{M}{\Delta t} \tilde{\mathbf{u}}_p^n - \sum_{j=1}^{s-1} a_{sj} \mathbf{R}(\tilde{\mathbf{u}}_p^{(j),n+1}) \right] = 0 \quad (s = 1, \dots, 6) \quad (12)$$

where \mathbf{R}_e^{n+1} represents the unsteady flow residual for time step $n + 1$. The current work uses a six-stage ESDIRK scheme (i.e. fourth-order accurate), and the values of the set of coefficients, a_{sj} , can be found in reference²⁷.

Since one or multiple non-linear problems are required to be solved in each implicit time step of the respective BDF2 scheme or the IRK4 scheme, in order to make these implicit schemes competitive, we employ efficient solution strategies including a non-linear p -multigrid approach driven by a linearized Gauss-Seidel smoother. Details about the solution methods can be found in references^{27,38}.

III. Adjoint-based Unsteady Shape Optimization

The adjoint sensitivity derivation starts with the formulation of the tangent linear problem, in which the discretized governing equations (c.f. Eq. (11) or Eq. (12)) are linearized and the sensitivity derivatives of an objective functional are formulated. The discrete adjoint formulation is then derived by transposing each matrix of the tangent linear problem and performing the operations in a reverse order. Here we first provide a preliminary description of the shape optimization techniques, and then we provide a detailed derivation for the calculation of sensitivity derivatives in the context of high-order discontinuous Galerkin discretizations on meshes involving curved boundary elements.

A. Preliminary Descriptions

1. Design variables and shape parametrization

Assume that an initial shape geometry, represented by the coordinates of boundary nodes \mathbf{x}_s and possibly additional surface quadrature points \mathbf{x}_q for curved elements, and a computational mesh are given. The surface geometry is modified through surface node displacements which are determined by a set of design variables, $\mathbf{D} = \{D_m, m = 1, \dots, N_d\}$ (where N_d represents the total number of design variables). The Hicks-Henne sine bump function³⁹ is employed in order to ensure smooth surface shapes, and the design variables are set to be the magnitudes of the bump functions placed at the surface nodes, expressed as:

$$b_i(x_{si}, D_m) = D_m \sin^4(\pi x_{si} \ln^{1/2} / \ln x_{sm}), \quad x_{si}, x_{sm} \in [0, 1] \quad (13)$$

where x_{sm} denotes the x -coordinate of the surface node where the bump function is placed. b_i denotes the surface node displacement at x_{si} due to the displacement of the surface node at x_{sm} , and D_m is the m^{th} component of the design variables associated with the surface node at x_{sm} . Fig. 2 depicts a sample plot of the bump functions with a magnitude

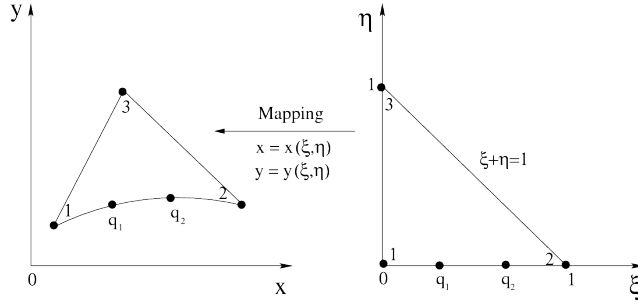


Figure 1. Geometric mapping between a physical curvilinear triangle and the reference triangle.

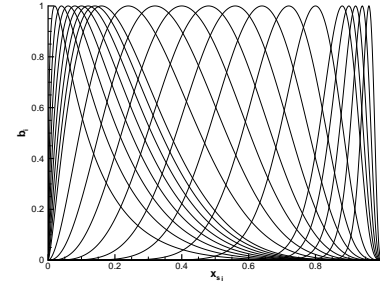


Figure 2. Illustration of the Hicks-Henne bump functions for various x_{sm} parameters varying from 0.02 to 0.96.

of 1, as x_{sm} takes on a range of 0.02 to 0.96. For multiple design variables ($N_d > 1$), the surface displacement at a particular node location of x_{si} is the superposition of all the bump functions placed at each designed surface node, and new surface coordinates are computed based on all the surface displacements, denoted as:

$$\begin{aligned} \Delta \mathbf{x}_{si} &= \mathbf{n}_{si} \sum_{m=1}^{N_d} b_i(x_{si}, D_m) \\ \mathbf{x}_{si}^{\text{new}} &= \mathbf{x}_{si}^{\text{old}} + \Delta \mathbf{x}_{si} \end{aligned} \quad (14)$$

where $\Delta \mathbf{x}_{si}$ represents the displacement at the surface node i . $\mathbf{n}_{si} = (n_{si_x}, n_{si_y})$ denotes the normal direction at the surface node i . $\mathbf{x}_{si}^{\text{(new)}}$ and $\mathbf{x}_{si}^{\text{(old)}}$ represent the new surface coordinates and the old ones from the initial design step, respectively. It is also noted that in many cases of aerodynamic airfoil design, surface deformation may occur only in the y-coordinate direction rather than in normal direction since it may be desirable to keep the chord length of the designed airfoil fixed. Then the resulting displacement direction vector denoted by \mathbf{n}_{si} in Eq. (14) is simply replaced by $(0, 1 \cdot \text{sign}(n_{si_y}))$ and thus the surface displacements in the x-coordinate direction vanish.

2. Deformation of additional quadrature points for surface shape

As described in Section II, the integrals in Eq. (5) are evaluated in the physical triangle but the basis set is defined in the reference triangle, thus a reference-to-physical coordinate transformation is required, as illustrated in Fig. 1. Additional quadrature points are required to determine higher-order geometric mapping modes $\tilde{\mathbf{x}}$ for curved surface elements (c.f. Eq. (9)), which must be employed in the presence of higher-order solutions ($p \geq 1$). Moreover, surface quadrature points must deform with surface nodes to define new surface mappings. This is achieved by using the same smooth bump functions used for the surface node displacements, shown as:

$$\begin{aligned} \Delta \mathbf{x}_{qi} &= \mathbf{n}_{qi} \sum_{m=1}^{N_d} b_i(x_{qi}, D_m) \\ \mathbf{x}_{qi}^{\text{new}} &= \mathbf{x}_{qi}^{\text{old}} + \Delta \mathbf{x}_{qi} \end{aligned} \quad (15)$$

where $\Delta \mathbf{x}_{qi}$ represents the displacement at the quadrature point i . $\mathbf{n}_{qi} = (n_{qi_x}, n_{qi_y})$ denotes its normal direction obtained by the old surface geometry, however it may be replaced by $(0, 1 \cdot \text{sign}(n_{qi_y}))$ to remain consistent with the surface node deformation. $\mathbf{x}_{qi}^{\text{(new)}}$ and $\mathbf{x}_{qi}^{\text{(old)}}$ represent the new surface quadrature coordinates for quadrature node i and the old ones from the initial design step, respectively. Therefore, as illustrated in Fig. 3, the displacements at the entire set of surface quadrature points are performed similarly to those occurring at the surface grid points.

3. Interior mesh deformation

Once the deformed surface mesh is obtained, the interior mesh is deformed to prevent generation of overlapping elements. Here we employ the linear tension spring analogy⁴⁰, in which each edge of the mesh is represented by a spring whose stiffness is related to the length of the edge. The governing equations are expressed as,

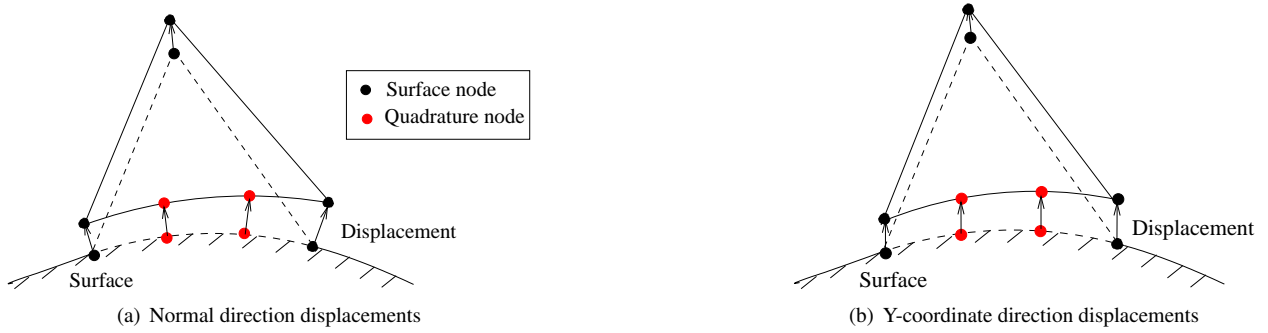


Figure 3. Illustration of deformation of curved surface elements with additional quadrature points. The dashed line element represents the initial element shape and the solid line element represents the deformed element shape.

$$[K]\Delta\mathbf{x} = \Delta\mathbf{x}_s \quad (16)$$

where $\Delta\mathbf{x}$ and $\Delta\mathbf{x}_s$ denote the entire mesh displacements and the surface mesh point displacements, respectively, and $[K]$ denotes the stiffness matrix obtained from the discrete mesh motion equations. This set of equations is solved using several hundred sweeps of a Gauss-Seidel scheme, since these equations are inexpensive to solve due to the relatively coarse meshes which are generally employed with high-order discretizations. Given the deformation of surface grid points, extra surface quadrature points and the interior mesh points, the corresponding coordinate mapping modal coefficients, $\tilde{\mathbf{x}}$, for the reference-to-physical mapping can be obtained, since the interior mesh contains only straight-sided elements (due to the inviscid nature of the problem being considered) and the use of element vertex coordinates is sufficient for determining the mapping.

4. Unsteady flow equations

The discretized implicit flow equations on the deformed mesh are expressed as:

$$\mathbf{R}_e \left(\tilde{\mathbf{u}}(\tilde{\mathbf{x}}), \tilde{\mathbf{u}}^0(\tilde{\mathbf{x}}), \mathbf{u}^b(\tilde{\mathbf{x}}), \tilde{\mathbf{x}} \right) = 0 \quad (17)$$

where $\mathbf{R}_e = \{\mathbf{R}_e^n, n = 1, 2, \dots, N\}$ represents the unsteady residual over all time steps and $\tilde{\mathbf{u}} = \{\tilde{\mathbf{u}}^n, n = 1, 2, \dots, N\}$ represents the unsteady flow solution by employing either the BDF2 scheme or the IRK4 scheme (N denotes the total number of time steps and the absorption of the subscript p is for simplicity). $\tilde{\mathbf{u}}^0$ and \mathbf{u}^b denote the initial condition and the unsteady free-stream boundary condition which may be explicitly dependent of the mesh geometry, such as in the case of the periodic gust response problem discussed in Section IV. Since the evaluation of the volume and surface integrals of the discretized governing equations (in Eq. (5)) requires the reference-to-physical transformation denoted by the coordinate mapping modal coefficients (in Eq. (8)), the discretized unsteady residual is shown as a function of $\tilde{\mathbf{x}}$ rather than \mathbf{x} , with the understanding that the mesh configuration (and thus $\tilde{\mathbf{x}}$) is held fixed throughout the flow time-integration marching and only changes between design iterations.

5. Objective functional

Consider a time-dependent scalar-valued functional, L , which refers to the objective functional for an optimization problem and is related to the time-dependent flow-field variables, such as time-integrated drag coefficients or target unsteady pressure distributions, etc. A general formulation for the objective functional is expressed as:

$$L = L(\tilde{\mathbf{x}}(\mathbf{D}), \tilde{\mathbf{u}}(\tilde{\mathbf{x}}(\mathbf{D}))) \quad (18)$$

where $\tilde{\mathbf{u}}$ denotes the computed unsteady flow solution and \mathbf{D} represents a set of input variables which produce changes in the surface mesh as described previously. This functional expression is written in terms of the coordinate modal coefficients rather than grid node coordinates for the reason described above. It is also shown that a variation in any component of \mathbf{D} not only produces changes in the element geometric mappings, but also produces changes in the time-dependent flow-field variables, which together contribute to a resulting variation in the objective functional.

B. Adjoint-based Sensitivity Calculation

The derivation of the unsteady discrete adjoint technique for sensitivity analysis starts with the forward tangent problem by taking the derivatives of Eq. (18) with respect to the design variables via the chain rule, which leads to the following expression for the sensitivity derivatives,

$$\frac{dL}{d\mathbf{D}} = \left(\frac{\partial L}{\partial \bar{\mathbf{x}}} + \frac{\partial L}{\partial \bar{\mathbf{u}}} \frac{\partial \bar{\mathbf{u}}}{\partial \bar{\mathbf{x}}} \right) \left(\frac{\partial \bar{\mathbf{x}}}{\partial \mathbf{x}} \frac{\partial \mathbf{x}}{\partial \mathbf{x}_s} \frac{\partial \mathbf{x}_s}{\partial \mathbf{D}} + \frac{\partial \bar{\mathbf{x}}}{\partial \mathbf{x}_q} \frac{\partial \mathbf{x}_q}{\partial \mathbf{D}} \right) \quad (19)$$

We next examine each term in the above equation. $\partial \mathbf{x}_s / \partial \mathbf{D}$ and $\partial \mathbf{x}_q / \partial \mathbf{D}$ refer to the sensitivity of surface mesh points and surface quadrature points with respect to the design variables, which are obtained by linearizing the definition of the bump functions as they appear in Equations (14) and (15), respectively. $\partial \mathbf{x} / \partial \mathbf{x}_s$ represents the sensitivity of the entire set of mesh points with respect to surface node displacements, which is evaluated based on the mesh deformation equation (16) as:

$$\frac{\partial \mathbf{x}}{\partial \mathbf{x}_s} = [\mathbf{K}]^{-1} \quad (20)$$

where $[\mathbf{K}]^{-1}$ represents the inverse of the mesh stiffness matrix. We note that the sensitivities $\partial \mathbf{x} / \partial \mathbf{x}_s$ rely purely on the grid node information, thus the stiffness matrix has the same formulation as the one derived in finite-volume methods⁴⁰. Due to the fact that the geometric mapping coefficients $\bar{\mathbf{x}}$ are determined by both mesh points \mathbf{x} and extra surface quadrature points \mathbf{x}_q , $\partial \bar{\mathbf{x}} / \partial \mathbf{x}$ and $\partial \bar{\mathbf{x}} / \partial \mathbf{x}_q$ provide the sensitivity of the coordinate mapping coefficients with respect to the entire set of mesh point displacements and surface quadrature point displacements, respectively. Moreover, the formulations in fact correspond to the relative entries of the inverse projection mapping matrix, Φ^{-1} defined in Eq. (9) in an element-wise manner. It is also noted that $\partial \bar{\mathbf{x}} / \partial \mathbf{x}_q$ has non-zero components only for physical boundary elements where high-order curved elements are allowed to occur. $\partial \bar{\mathbf{u}} / \partial \bar{\mathbf{x}}$ denotes the flow sensitivities due to perturbations of the geometric mapping coefficients, $\bar{\mathbf{x}}$, which result from the mesh deformation, and can be formulated by linearizing Eq. (17) with respect to the design variables via the chain rule, as:

$$\left(\left[\frac{\partial \mathbf{R}_e}{\partial \bar{\mathbf{u}}} \right] \frac{\partial \bar{\mathbf{u}}}{\partial \bar{\mathbf{x}}} + \left[\frac{\partial \mathbf{R}_e}{\partial \bar{\mathbf{u}}^0} \right] \frac{\partial \bar{\mathbf{u}}^0}{\partial \bar{\mathbf{x}}} + \left[\frac{\partial \mathbf{R}_e}{\partial \mathbf{u}^b} \right] \frac{\partial \mathbf{u}^b}{\partial \bar{\mathbf{x}}} + \frac{\partial \mathbf{R}_e}{\partial \bar{\mathbf{x}}} \right) \frac{\partial \bar{\mathbf{x}}}{\partial \mathbf{D}} = 0 \quad \text{or} \quad \frac{\partial \bar{\mathbf{u}}}{\partial \bar{\mathbf{x}}} = - \left[\frac{\partial \mathbf{R}_e}{\partial \bar{\mathbf{u}}} \right]^{-1} \frac{\partial \bar{\mathbf{R}}_e}{\partial \bar{\mathbf{x}}} \quad (21)$$

where

$$\frac{\partial \bar{\mathbf{R}}_e}{\partial \bar{\mathbf{x}}} = \frac{\partial \mathbf{R}_e}{\partial \bar{\mathbf{x}}} + \left[\frac{\partial \mathbf{R}_e}{\partial \bar{\mathbf{u}}^0} \right] \frac{\partial \bar{\mathbf{u}}^0}{\partial \bar{\mathbf{x}}} + \left[\frac{\partial \mathbf{R}_e}{\partial \mathbf{u}^b} \right] \frac{\partial \mathbf{u}^b}{\partial \bar{\mathbf{x}}} \quad (22)$$

It is noted that the sensitivity of the unsteady flow residual with respect to the flow initial condition has contributions only from $[\partial \mathbf{R}_e^1 / \partial \bar{\mathbf{u}}^0]$ and $[\partial \mathbf{R}_e^2 / \partial \bar{\mathbf{u}}^0]$ in the BDF2 scheme, and from $[\partial \mathbf{R}_e^{(s),1} / \partial \bar{\mathbf{u}}^0]$ for the intermediate stages of the first time step in the IRK4 scheme. However, the last two terms in Eq. (22) vanish if uniform initial and free-stream boundary conditions are given. Substituting Eq. (20) and Eq. (21) into Eq. (19) yields the expression for the gradient sensitivities, shown as:

$$\frac{dL}{d\mathbf{D}} = \left(\frac{\partial L}{\partial \bar{\mathbf{x}}} - \frac{\partial L}{\partial \bar{\mathbf{u}}} \left[\frac{\partial \mathbf{R}_e}{\partial \bar{\mathbf{u}}} \right]^{-1} \frac{\partial \bar{\mathbf{R}}_e}{\partial \bar{\mathbf{x}}} \right) \left(\frac{\partial \bar{\mathbf{x}}}{\partial \mathbf{x}} [\mathbf{K}]^{-1} \frac{\partial \mathbf{x}_s}{\partial \mathbf{D}} + \frac{\partial \bar{\mathbf{x}}}{\partial \mathbf{x}_q} \frac{\partial \mathbf{x}_q}{\partial \mathbf{D}} \right) \quad (23)$$

The discrete adjoint problem is then formulated by transposing both sides of Eq. (23), which yields:

$$\frac{dL^T}{d\mathbf{D}} = \left(\frac{\partial \mathbf{x}_s^T}{\partial \mathbf{D}} [\mathbf{K}]^{-T} \frac{\partial \bar{\mathbf{x}}^T}{\partial \mathbf{x}} + \frac{\partial \mathbf{x}_q^T}{\partial \mathbf{D}} \frac{\partial \bar{\mathbf{x}}^T}{\partial \mathbf{x}_q} \right) \left(\frac{\partial L^T}{\partial \bar{\mathbf{x}}} - \frac{\partial \bar{\mathbf{R}}_e^T}{\partial \bar{\mathbf{x}}} \left[\frac{\partial \mathbf{R}_e}{\partial \bar{\mathbf{u}}} \right]^{-T} \frac{\partial L^T}{\partial \bar{\mathbf{u}}} \right) \quad (24)$$

where $[\cdot]^{-T}$ denotes the inverse of the transposed matrix. Eq. (24) corresponds to the formulation for evaluating the discrete adjoint-based sensitivity derivatives in the context of discontinuous Galerkin discretizations. Some notable differences between the above formulation for DG methods and the corresponding adjoint formulation¹⁷ in finite-volume methods can be observed: first, both surface grid points and surface quadrature points are deformed simultaneously based on the bump functions: the consequent sensitivity terms relevant to the design variables in this formulation include both $(\partial \mathbf{x}_s / \partial \mathbf{D})^T$ and $(\partial \mathbf{x}_q / \partial \mathbf{D})^T$. Second, the sensitivities of the coordinate mapping, $(\partial \bar{\mathbf{x}} / \partial \mathbf{x})^T$ and $(\partial \bar{\mathbf{x}} / \partial \mathbf{x}_q)^T$, are required to be evaluated accordingly. As discussed previously, these terms represent the effect of the entire set of mesh points and the surface quadrature points on the geometric mapping coefficients. Third, the

sensitivities denoted in the second bracket of the right-hand-side of the equation, such as $[\partial \mathbf{R}_e / \partial \tilde{\mathbf{x}}]^T$ or $(\partial L / \partial \tilde{\mathbf{x}})^T$ must be evaluated with respect to the corresponding modal coefficients of geometric mappings in the DG discretizations, as opposed to the direct use of grid coordinates in finite-volume methods.

We note that the term $[\partial \mathbf{R}_e / \partial \tilde{\mathbf{u}}]$ in Eq. (24) corresponds to the Jacobian matrix of the full unsteady residual. Since a direct solve for the inverse of the full Jacobian matrix or its transpose can be very expensive, the unsteady flow-adjoint variables, λ_u , are introduced by replacing the last two terms, satisfying:

$$\left[\frac{\partial \mathbf{R}_e}{\partial \tilde{\mathbf{u}}} \right]^{-T} \frac{\partial L^T}{\partial \tilde{\mathbf{u}}} = \lambda_u \quad \text{or} \quad \left[\frac{\partial \mathbf{R}_e}{\partial \tilde{\mathbf{u}}} \right]^T \lambda_u = \frac{\partial L^T}{\partial \tilde{\mathbf{u}}} \quad (25)$$

Thus the transpose of the Jacobian of the discrete unsteady flow equations is used in the definition of the flow-adjoint variables, evaluated using the computed unsteady flow states, $\{\tilde{\mathbf{u}}^n\}$. Therefore, the unsteady flow-adjoint solution corresponds to a series of linear problems spanning the entire time domain, expressed as $\{\lambda_u^n\}$. Details about the solution procedure of the unsteady flow-adjoint problems for the implicit temporal schemes employed in this work are discussed in the next section. Substituting the unsteady flow-adjoint variables λ_u into Eq. (24) yields:

$$\frac{dL^T}{d\mathbf{D}} = \frac{\partial \mathbf{x}_s^T}{\partial \mathbf{D}} [K]^{-T} \frac{\partial \bar{L}^T}{\partial \tilde{\mathbf{x}}} + \frac{\partial \mathbf{x}_q^T}{\partial \mathbf{D}} \frac{\partial \bar{L}^T}{\partial \mathbf{x}_q} \quad (26)$$

where

$$\frac{\partial \bar{L}^T}{\partial \tilde{\mathbf{x}}} = \frac{\partial \tilde{\mathbf{x}}^T}{\partial \tilde{\mathbf{x}}} \frac{\partial L^T}{\partial \tilde{\mathbf{x}}} \quad (27)$$

$$\frac{\partial \bar{L}^T}{\partial \mathbf{x}_q} = \frac{\partial \tilde{\mathbf{x}}^T}{\partial \mathbf{x}_q} \frac{\partial L^T}{\partial \tilde{\mathbf{x}}} \quad (28)$$

$$\frac{\partial \bar{L}^T}{\partial \tilde{\mathbf{x}}} = \frac{\partial L^T}{\partial \tilde{\mathbf{x}}} - \frac{\partial \mathbf{R}_e^T}{\partial \tilde{\mathbf{x}}} \lambda_u \quad (29)$$

where $(\partial \bar{L} / \partial \tilde{\mathbf{x}})^T$ denotes objective sensitivities with respect to the mesh deformation and $(\partial \bar{L} / \partial \mathbf{x}_q)^T$ denotes objective sensitivities with respect to the surface quadrature point deformation.

Returning to Eq. (26), a similar approach is taken to avoid a direct solve for the inverse of the transposed mesh stiffness matrix $[K]^{-T}$ by introducing the mesh-adjoint variables, λ_x , satisfying:

$$[K]^{-T} \frac{\partial \bar{L}^T}{\partial \tilde{\mathbf{x}}} = \lambda_x \quad \text{or} \quad [K]^T \lambda_x = \frac{\partial \bar{L}^T}{\partial \tilde{\mathbf{x}}} \quad (30)$$

Substituting Eq. (30) to Eq. (26), the final expression for the sensitivity derivatives is then shown as:

$$\frac{dL^T}{d\mathbf{D}} = \frac{\partial \mathbf{x}_s^T}{\partial \mathbf{D}} \lambda_x + \frac{\partial \mathbf{x}_q^T}{\partial \mathbf{D}} \frac{\partial \bar{L}^T}{\partial \mathbf{x}_q} \quad \text{or} \quad (31)$$

$$\frac{dL^T}{d\mathbf{D}} = \lambda_x^T \frac{\partial \mathbf{x}_s}{\partial \mathbf{D}} + \frac{\partial \bar{L}}{\partial \mathbf{x}_q} \frac{\partial \mathbf{x}_q}{\partial \mathbf{D}} \quad (32)$$

Since the terms relevant to the design variables (i.e. $\frac{\partial \mathbf{x}_s^T}{\partial \mathbf{D}}$ and $\frac{\partial \mathbf{x}_q^T}{\partial \mathbf{D}}$) are evaluated at the last step in this formulation, the evaluation of the adjoint-based sensitivity derivatives is essentially independent of the number of design variables, which makes the adjoint method well-suited for cases with large numbers of design variables.

C. Unsteady Flow-adjoint Formulation

In this section, we focus on the derivation of the unsteady flow-adjoint formulation denoted in Eq. (25) for the fourth-order implicit Runge-Kutta scheme, and we refer readers to the work^{38,41} for a general backward difference temporal scheme.

where $(\partial L / \partial \bar{\mathbf{u}}^{(s),n})^T$ shows the dependence of the selected objective functional on an intermediate-stage flow solution ($s < 6$), which turns out to be zero if the objective is a function of the solutions at only the discrete time step locations. It is noted that the right-hand-side of the adjoint formulation in Eq. (34) for evaluating the adjoint problem at the final time step ($s = 6, n = N$) is different from that in Eq. (36) for the evaluation at an intermediate time step ($s = 6, n \neq N$). In addition, the flow Jacobian matrices³⁸ denoted in Equations (34)-(37) can be evaluated by linearizing Eq. (12) with respect to the appropriate flow state, and the solution strategy for solving each of the linear problems uses the p -multigrid approach³⁸ to accelerate the solution convergence. Given the flow-adjoint solution, the vector denoted by the multiplication $[\partial \bar{\mathbf{R}}_e / \partial \tilde{\mathbf{x}}]^T \lambda_u$ in Eq. (29) is then calculated by the summation $\sum_{n=1}^N \sum_{s=2}^6 [\partial \bar{\mathbf{R}}_e^{(s),n} / \partial \tilde{\mathbf{x}}]_u^T \lambda_u^{(s),n}$.

D. Shape Optimization Procedure

Once the objective sensitivities have been evaluated, they are used to drive a design optimization process to seek a minimum in a specified objective functional. It is well known that the efficiency of an optimization process relates closely to the particular optimization algorithm. While a gradient-based steepest descent method is simple to implement, this approach tends to deliver slow convergence when a large number of design variables are employed. The current work employs the PORT trust region optimization strategy⁴², which chooses a region around the current input variables at each iteration to obtain a new set of inputs. In addition, PORT requires the initial bounds to be specified for the set of design variables at the first design iteration to seek the trust region and the bounds are generally selected to be reasonable for the analysis solver and to avoid generation of collapsed shapes.

For a typical design-optimization cycle, five sequential steps are required:

1. Solve the unsteady flow equations using either the BDF2 scheme or the IRK4 scheme.
2. Solve the unsteady flow-adjoint variables and then the mesh adjoint variables denoted in Eq. (25) and Eq. (30) respectively.
3. Evaluate gradients or objective functional sensitivities $dL/d\mathbf{D}$ described in the previous section (c.f. Eq. (31) or Eq. (32)).
4. Compute a new set of design variables, \mathbf{D}^{new} , using the PORT optimization algorithm based on the computed sensitivity derivatives.
5. Deform the geometry shape including additional surface points for curved elements using the new design variables as the magnitudes of the Hicks-Henne bump function and then deform the interior mesh based on the mesh motion equations.

This procedure is repeated until the objective functional is sufficiently minimized.

IV. Numerical Results

In this section, we first investigate the applicability of the current nonlinear implicit DG solver for the single airfoil aeroacoustics gust response problem²⁸ via comparisons of the unsteady surface pressure disturbances as well as acoustic intensity in the model problem with those obtained from NASA BASS and GUST3D computational aeroacoustics codes^{28,31,43}. Next two numerical examples of unsteady airfoil shape optimization are presented to demonstrate the performance of the adjoint-based shape optimization strategy, where the first aims to change the shape of the original NACA0012 airfoil to match a time-dependent surface pressure profile produced by a two-dimensional vortical gust impinging on a RAE-2822 airfoil, and the objective of the second is to minimize the acoustic noise generated by subsonic flow over a NACA0012 airfoil with a $0.03c$ blunt trailing edge.

A. Analysis Study for Single Airfoil Gust Response Problem

Consider a two-dimensional vortical gust which convects past a 12% thick, zero-degree angle of attack Joukowski airfoil with the gust velocity distribution given by,

$$u_g = -(\epsilon k_2 M_\infty a_\infty / \sqrt{k_1^2 + k_2^2}) \cos(k_1 x + k_2 y - \omega t) \quad (38)$$

$$v_g = (\epsilon k_1 M_\infty a_\infty / \sqrt{k_1^2 + k_2^2}) \cos(k_1 x + k_2 y - \omega t) \quad (39)$$

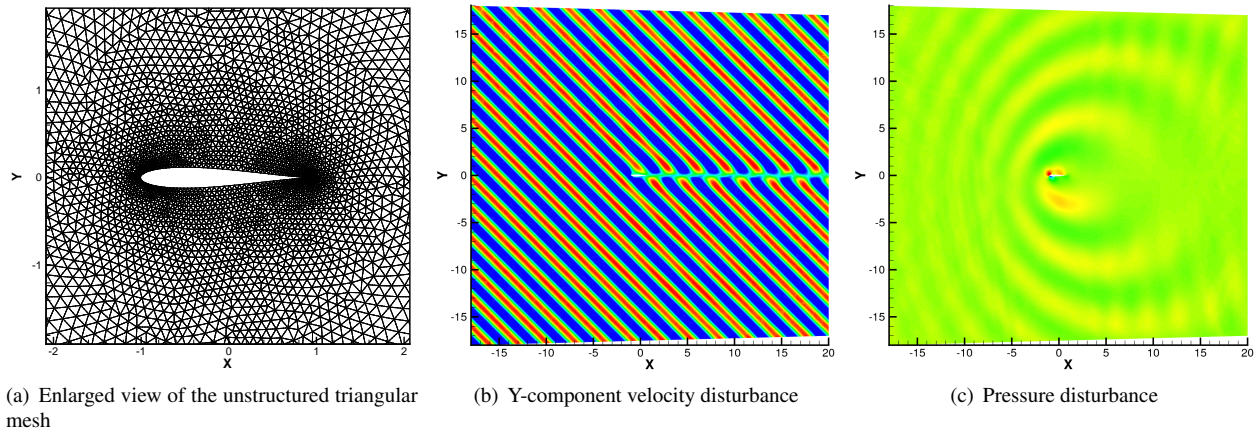


Figure 4. Computational mesh ($N = 13739$) and instantaneous solution contours for a two-dimensional gust over a symmetric Joukowski airfoil with a reduced frequency of $k_1 = 2$, using a fourth-order spatial DG discretization and the implicit BDF2 temporal scheme.

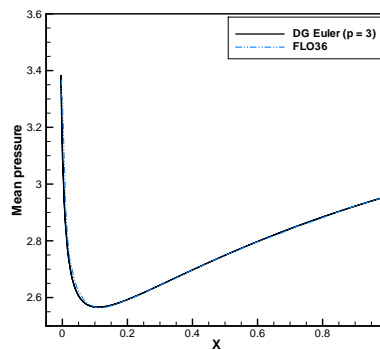


Figure 5. Comparison of DG Euler solution with potential flow solution from FLO36⁴⁵ for the mean pressure distribution on the symmetric Joukowski airfoil in the gust response test case.

in the upstream, where $\mathbf{k} = (k_1, k_2)$ denotes the wave number vector which specifies the direction of the gust propagation. ϵ is a small parameter satisfying $\epsilon \ll 1$. M_∞ and a_∞ denote the Mach number and speed of sound in the free-stream, respectively. ω represents the gust angular frequency, which is equal to $2k_1 M_\infty a_\infty / c$, where c denotes the airfoil chord length. k_1 is set equal to k_2 and denotes the reduced frequency. The mean flow parameters at infinity are specified as $(\rho_\infty, u_\infty, v_\infty, p_\infty) = (\gamma, M_\infty, 0, 1)$ where $M_\infty = 0.5$.

A fourth-order accurate (i.e. $p = 3$) DG discretization and the implicit BDF2 scheme with a time-step size of 0.001 are employed for the respective spatial and temporal discretizations. The HLLC approximate Riemann solver³² is used for the flux function at all interior edges/boundaries and the particular wall boundary treatment described in Section II is employed at all wall/surface boundaries. The unsteady flow-field simulation starts with a steady-state solution in the absence of the gust and then the time-dependent gust problem is solved until the flow becomes periodic.

The RMS pressure disturbance $\sqrt{\overline{(p')^2}}$ on the airfoil surface and the acoustic intensity $\overline{(p')^2}$ at circular locations of $R = 1c$, $R = 2c$ and $R = 4c$ are calculated for three reduced frequencies consisting of $k_1 = 0.1$, $k_1 = 1$ and $k_1 = 2$. Current results are compared with the GUST3D linearized Euler solution and the high-order nonlinear BASS solution documented in references^{28,31,44}.

Fig. 4 (a) shows the computational mesh used in the gust problem which contains 13739 unstructured triangular elements. Figures 4 (b)-(c) illustrate representative flow-field results for the instantaneous y-component velocity disturbance and the instantaneous pressure disturbance in the case of a two-dimensional gust impinging on the symmetric Joukowski airfoil at a reduced frequency of $k_1 = 2$. It is seen that the periodic gust starts to disturb the flow field from upstream and convects downstream. Fig. 4 (c) exhibits the qualitative features of the acoustic field noise, generated by

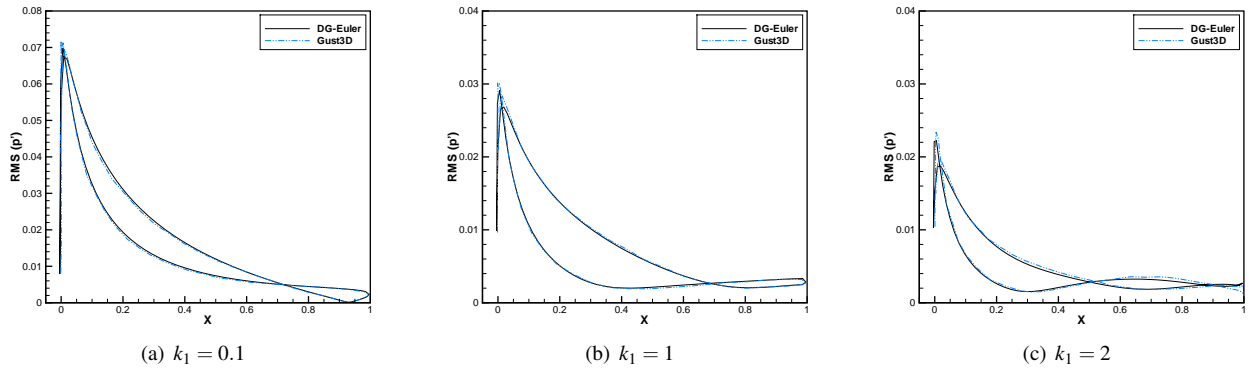


Figure 6. Comparison of DG Euler solution with GUST3D²⁸ solution for the RMS pressure disturbance distribution on the symmetric Joukowski airfoil surface in the gust response test case.

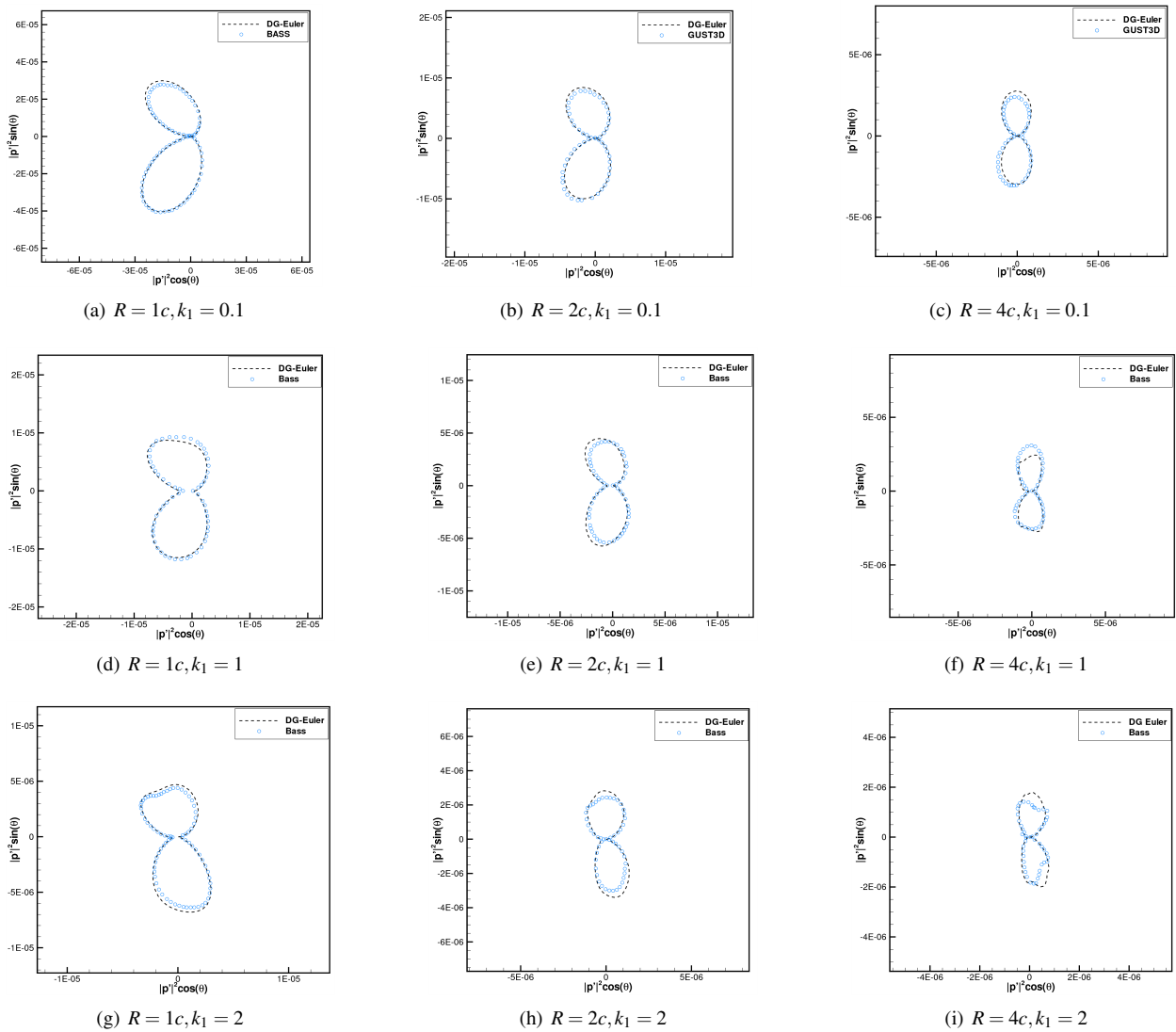


Figure 7. Comparison of DG Euler solution with GUST3D²⁸ and BASS³¹ solutions for the acoustic intensity distribution at locations of $R = 1c, R = 2c$ and $R = 4c$ in the gust response test case with reduced frequencies of $k_1 = 0.1, k_1 = 1$ and $k_1 = 2$.

the interaction of the gust with the Joukowski airfoil. The mean pressure distribution on the airfoil surface is plotted in Fig. 5 and is compared with the FLO36 potential flow solution⁴⁵ digitized from reference⁴⁴, where good agreement can be observed. The corresponding RMS pressure disturbances on the airfoil surface are shown in Fig. 6 for various reduced frequencies, and are compared to the GUST3D linearized Euler solution²⁸, where the agreement of the current DG-Euler solution with the GUST3D benchmark solution is very good. In Fig. 7, a series of computed results on acoustic intensity at different far-field locations away from the centerpoint of the airfoil are compared with the high-order nonlinear BASS solution and/or the GUST3D solution for various reduced frequencies. One can see that the current DG Euler results consistently agree well in most of the cases, although a significantly finer mesh (with about ninety thousand mesh points) was used in the BASS simulations. Only minor discrepancies are noted in the case of $R = 4c$ for the reduced frequency of $k_1 = 2$, but the difference is small.

B. Unsteady Shape Optimization for the Single Airfoil Gust Response Problem

The first design problem used to demonstrate the performance of the adjoint-based shape optimization considers the single airfoil vorticity gust response problem discussed previously, where the root mean square difference between pressure distributions on the target RAE-2822 airfoil configuration with the original NACA0012 airfoil configuration is selected as the objective functional, given by,

$$L = \sqrt{\frac{\sum_{n=n_s}^N \sum_{j=1}^{N_s} \sum_{q=1}^{N_q} \left(p_{q,j}^n - (p_{q,j}^n)^* \right)^2}{N^* \cdot N_s \cdot N_q}} \quad (40)$$

where $p_{q,j}^n$ represents the pressure value obtained from the current airfoil configuration at the n^{th} time step for quadrature point q at the airfoil surface edge j , and $(p_{q,j}^n)^*$ represents the pressure value for the target RAE-2822 airfoil configuration at the same location and time step. The respective n_s , N and N^* refer to the starting and ending time-step indices and the number of time steps in the calculation of the objective functional. N_s and N_q denote the total number of surface edges and quadrature points per edge, respectively. Note that the target pressure distribution is a function of airfoil geometry curvature. In order to assure a target pressure value at a specified location is deformably achievable by the original airfoil, the pressure values initially obtained from the target airfoil must be interpolated to obtain a set of pressure targets which correspond to the locations of the original airfoil in the functional definition for each time step. Furthermore, the objective functional of the RMS difference is theoretically capable of reaching a smaller value as the interpolation error is decreased. In this test case, a cubic interpolation function is used to obtain the target pressure values according to the x-coordinates of the quadrature points for the original NACA0012 computational mesh, and thus the displacements of surface grid points as well as additional surface quadrature points are only allowed in the y-coordinate direction for this interpolation approach. The design variables are set to be the magnitudes of the bump functions placed at surface grid points, ranging from 2% to 99% of the chord locations of the upper and lower airfoil surfaces. Once the objective functional and design variables are given, the sensitivity gradients can be evaluated using the discrete adjoint sensitivity technique described previously.

The initial geometry (NACA0012) and the computational mesh, which contains 4657 unstructured triangular elements and 117 design variables, are illustrated in Fig. 8 (a). A mean flow Mach number of 0.5 and a zero-degree angle of attack are specified for this design problem. The inflow vortical gust has a distribution as described previously and we set the reduced frequency to be 5 in this case. A fourth-order (i.e. $p = 3$) spatial discontinuous Galerkin scheme and the fourth-order implicit Runge-Kutta (IRK4) scheme are employed for the respective spatial and temporal discretizations. A steady-state solution for the original NACA0012 airfoil is solved in the absence of the gust and is used as the initial flow condition for the design optimization problem. The time-step size and a total number of time steps employed are equal to 0.1 and 150, respectively. Fig. 8 (b) illustrates the instantaneous x-component velocity disturbance at the final time $T = 15$ in the case of a vortical gust impinging on the original NACA0012 airfoil with a reduced frequency of $k_1 = 5$. Furthermore, in order to eliminate flow transition effects to the shape optimization problem, only the last 25 time steps, which correspond to roughly one period of the periodic flow solution, are selected for the time interval in which the objective functional is computed. Fig. 9 depicts the time-dependent lift coefficient profile on the original airfoil configuration for the gust problem, where the red box indicates the actual objective time-interval. Additionally, the flow and adjoint problems in the design procedure are solved using a p -multigrid approach²⁷ with 10 smoothing passes at each multigrid level.

Before proceeding to the design optimization procedure, the sensitivity vector computed using the proposed discrete adjoint procedure is verified by comparing with the finite-differenced gradients, where the original airfoil geometry (NACA0012) and computational mesh are employed in this validation test. In the finite difference scheme, each

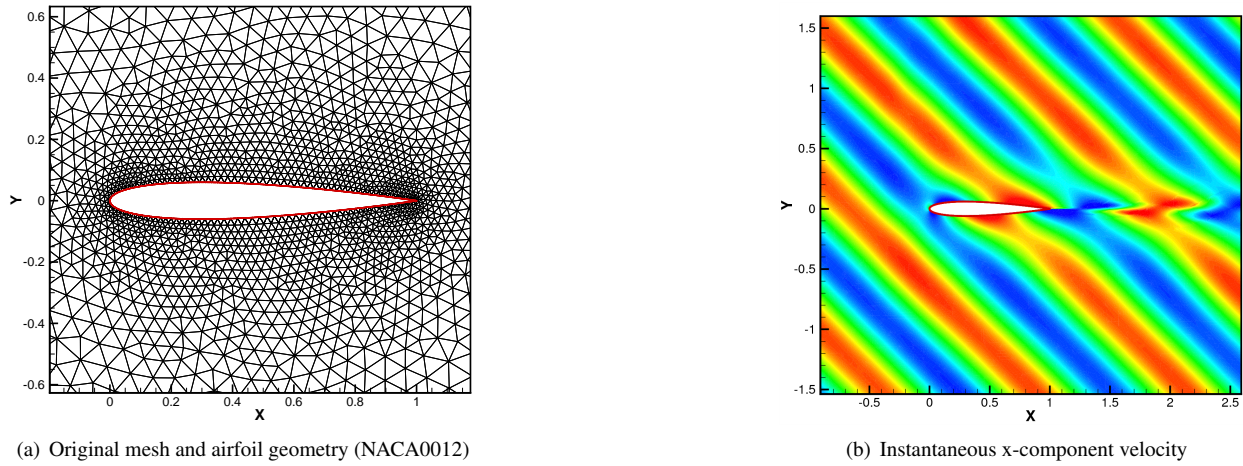


Figure 8. (a) Original computational mesh and the NACA0012 airfoil geometry. (b) Contours of the instantaneous x-component velocity for a two-dimensional vorticity gust over the original NACA0012 airfoil with a reduced frequency of $k = 5$ at final time $T = 15$, using a fourth-order discontinuous Galerkin discretization and the implicit IRK4 scheme.

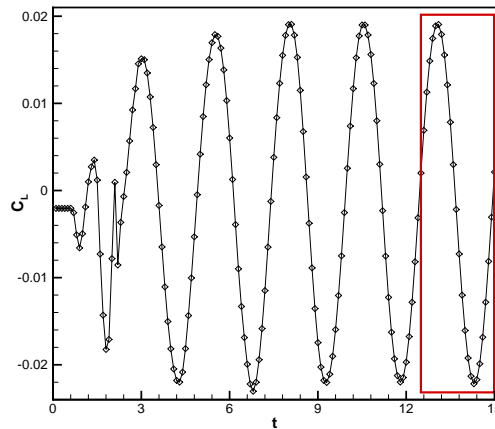


Figure 9. Profile of time-dependent lift coefficient for the original NACA0012 airfoil in the inverse shape optimization test case, where the red box indicates the time interval in which the objective functional is computed.

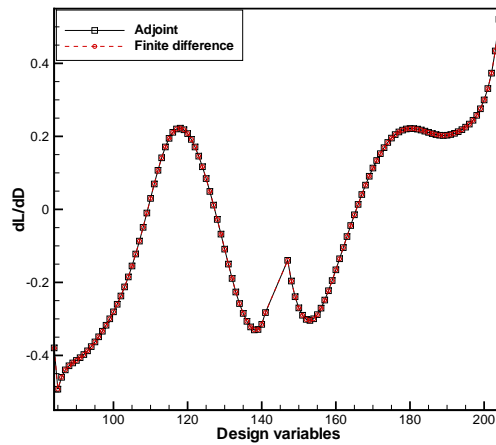


Figure 10. Comparison of sensitivity derivatives using the unsteady discrete adjoint method and the finite-difference method for the original mesh and airfoil geometry (NACA0012).

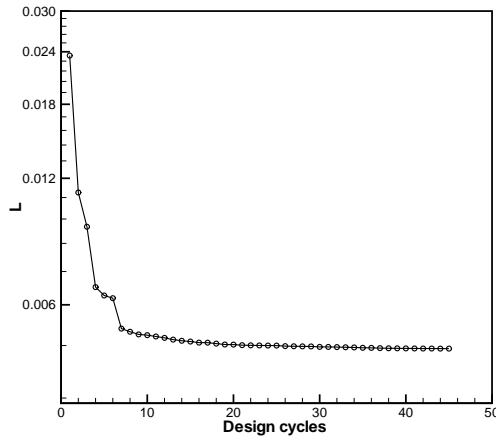


Figure 11. Convergence of the objective functional in terms of the number of design cycles in the gust response shape optimization test case.

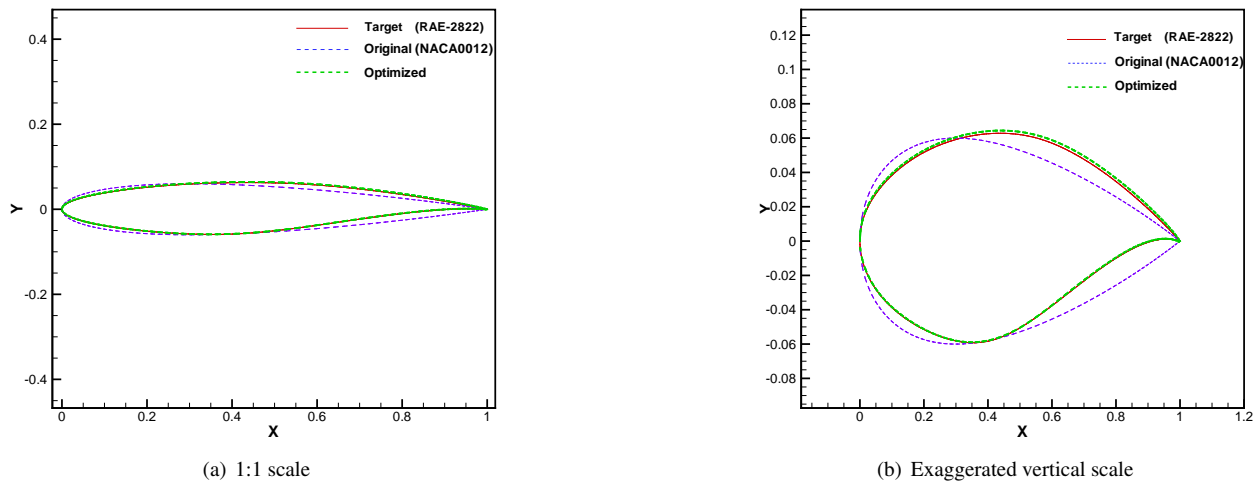
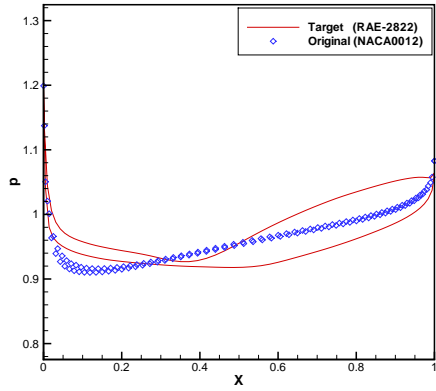


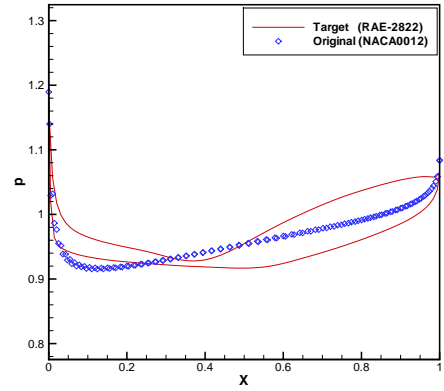
Figure 12. Comparison of airfoil shapes among the target RAE-2822 airfoil, the original NACA0012 airfoil and the final optimized airfoil in the gust response shape optimization test case.

component of the sensitivity vector (e.g. dL/dD_i) is obtained by perturbing the bump function placed at the airfoil surface point i , obtaining a new curved surface geometry and deformed computational mesh, and then re-computing the unsteady flow solution to obtain the change of the objective functional caused by the perturbation of D_i . This procedure is invoked for all design variables to obtain the entire sensitivity vector. Fig. 10 illustrates the comparison between the sensitivity values computed using the discrete adjoint approach for the $p = 3$ DG discretization and the IRK4 temporal scheme with those obtained from the finite-difference scheme. As the design variable index increases, the values shown in this figure are obtained from the original lower airfoil surface starting from the trailing edge to the leading edge, and then from the upper airfoil surface starting from the leading edge and ending at the trailing edge. It is shown that the computed adjoint sensitivities provide an excellent match with the finite differenced results, and the average difference between the two approaches is about 10^{-5} , thus verifying the linearization terms formulated in Eq. (24).

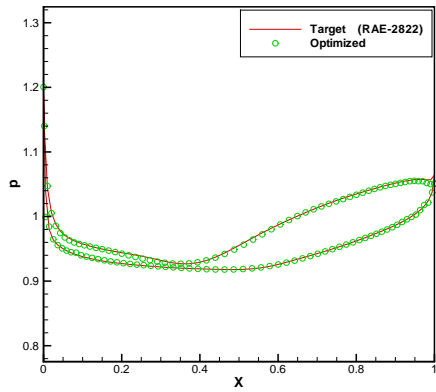
Fig. 11 plots the computed objective functional against the number of shape optimization cycles, where it is shown that the objective functional decreases significantly for the initial design steps. The optimization process stops after 45 design cycles in that only subtle changes to the shape are produced beyond that point, however, an approximate one order of magnitude reduction in the objective is achieved in 45 design iterations. Fig. 12 compares the geometry shapes for the original, target and final optimized airfoils using a 1:1 scale as well as an exaggerated vertical scale to show details. It is evident that the target RAE-2822 airfoil shape differs considerably from the original symmetric



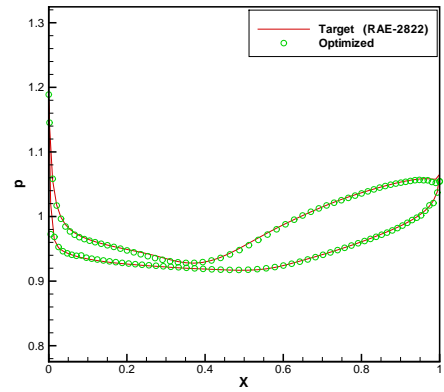
(a) Design cycle 0, time step 142



(b) Design cycle 0, time step 150

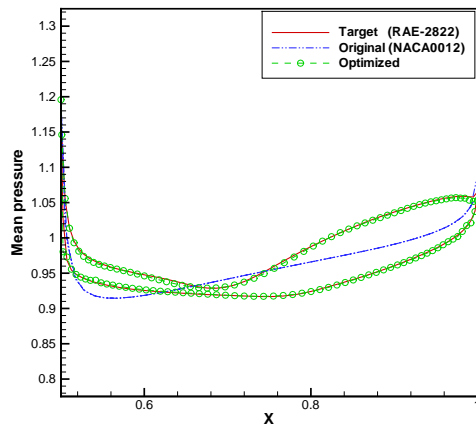


(c) Design cycle 45, time step 142

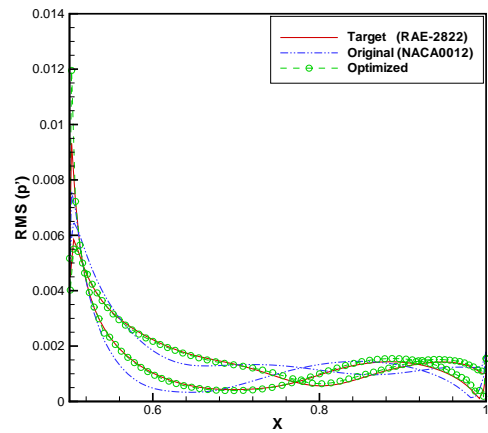


(d) Design cycle 45, time step 150

Figure 13. Comparison of pressure distributions on the airfoil surfaces at various time steps for the original and the final optimized airfoils for the gust response shape optimization test case.



(a) Mean pressure



(b) RMS pressure disturbance

Figure 14. Comparison of mean pressure distributions and RMS pressure disturbances on the target (RAE-2822), the original (NACA0012) and the final optimized airfoil surfaces in the gust response shape optimization test case.

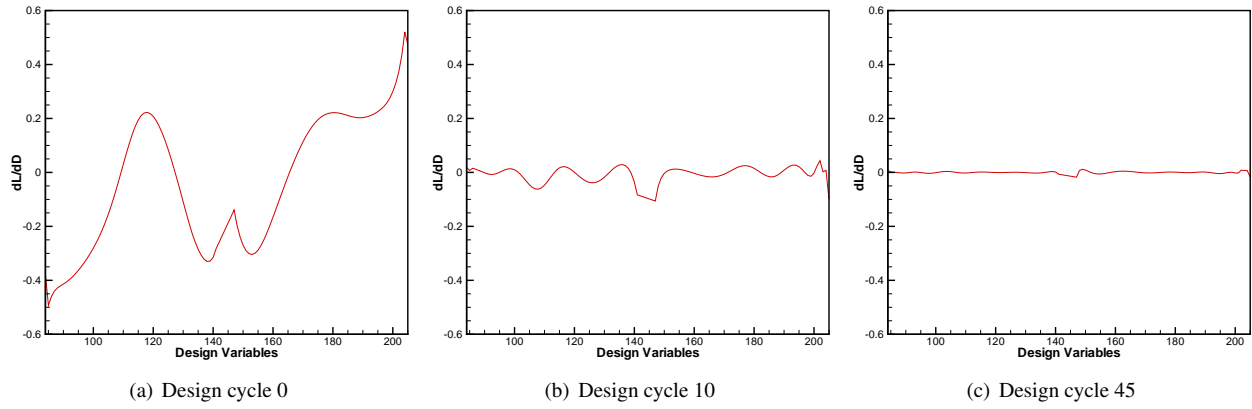


Figure 15. Distribution of adjoint-based sensitivity derivatives at various design cycles in the gust response shape optimization test case, where the gradient norms are 0.2516, 0.0264 and 0.0037, respectively.

NACA0012 airfoil, especially near the lower trailing edge. However, as illustrated in Figures 12 (a)-(b), the consequent optimized shape matches the target very well and there are no visible discrepancies on the lower airfoil surfaces even in the exaggerated scale. In Fig. 13, a comparison of the time-dependent pressure profile on the original, target and optimized airfoils is provided for various time steps at the initial and final design cycles. It is seen that the airfoil geometry has a strong effect on the airfoil unsteady pressure in the gust response problem, however, the target unsteady pressure profile is captured very accurately by the final optimized airfoil. Fig. 14 exhibits comparison of the mean pressure and RMS pressure disturbance distributions on these airfoils obtained from a one-period time integration. It is shown that the optimized airfoil consistently provides good agreement on the target mean pressure and pressure fluctuation profiles, while only a slight overprediction in the RMS pressure disturbance is observed near the upper leading edge. The distributions of the adjoint-based sensitivity derivatives at various design cycles are depicted in Fig. 15, where the discontinuity in the sensitivity distributions at design variable indices 141 and 147 implies the transition from the lower airfoil surface to the upper surface. Due to the rapid objective convergence in the initial design steps, the norm of the sensitivity derivatives is decreased by about one order of magnitude within only 10 design iterations. In addition, a two-order magnitude reduction in gradient norm is approximately obtained in 45 design iterations, as seen in Fig. 15 (c) where the sensitivities are all driven to be considerably small at the final design cycle.

C. Airfoil Shape Design for Blunt Trailing Edge Acoustic Radiation

The second test case considers pressure fluctuation (noise) minimization for a blunt trailing edge flow⁴⁶. Fig. 16 illustrates the original computational mesh which consists of 4942 elements and 100 surface points for a NACA0012 airfoil with a $0.03c$ ($H=0.03$) thick blunt trailing edge and a chopped chord length of 0.8894. A free-stream Mach number of 0.2 and a zero-degree angle of attack are specified in this case. Due to the sharp corners of the blunt trailing edge, vortices are generated, separate and are then convected downstream, producing pressure fluctuations in near-to-far fields from the airfoil body. In this case, we seek to minimize the pressure fluctuations (i.e. the difference between instantaneous pressure and mean pressure) at the one-chord length circular locations from the centerpoint of the NACA0012 blunt trailing edge airfoil, as indicated by the red circle in Fig. 16 (a). Therefore the objective functional for this purpose is given by,

$$L = \sqrt{\frac{\sum_{n=n_s}^N \sum_{q=1}^{N_p} (p_q^n - \bar{p}_q)^2}{N^* \cdot N_p}} \quad (41)$$

where N_p denotes the total number of grid points used to evaluate the noise generation at the one-chord length locations. n_s and N refer to the starting and ending time-step indices respectively. p_q^n represents the unsteady pressure at grid point q and time step n and \bar{p}_q denotes the mean values of the unsteady pressure at the corresponding location, which is calculated as,

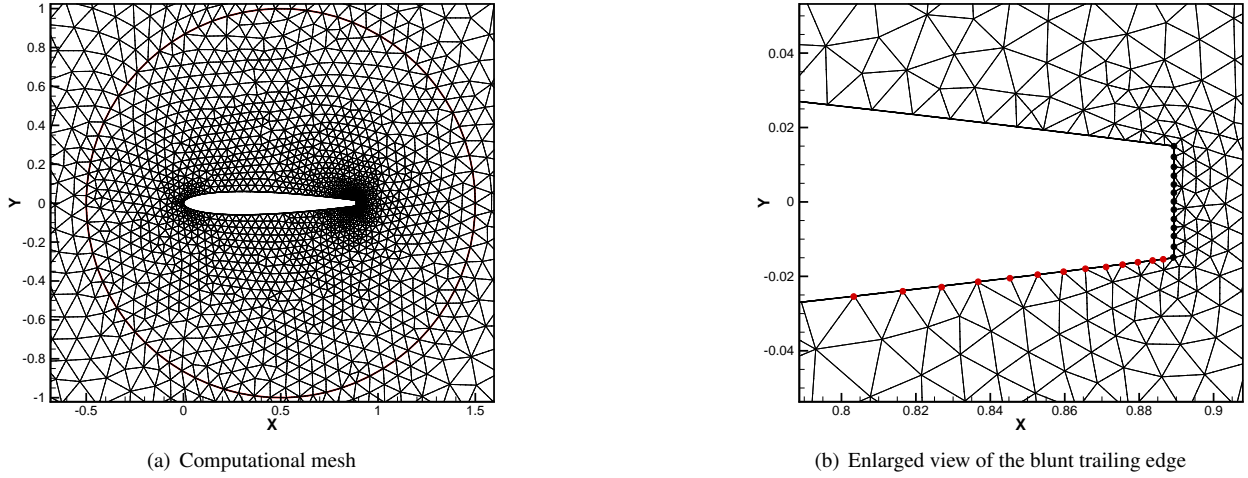


Figure 16. Computational mesh and surface geometry of the initial NACA0012 airfoil with a $0.03c$ blunt trailing edge. The red circle in the left figure indicates areas where pressure fluctuations are required to be minimized, and red dots in the right figure indicate locations of the design variables, while the black dots indicate fixed grid points.

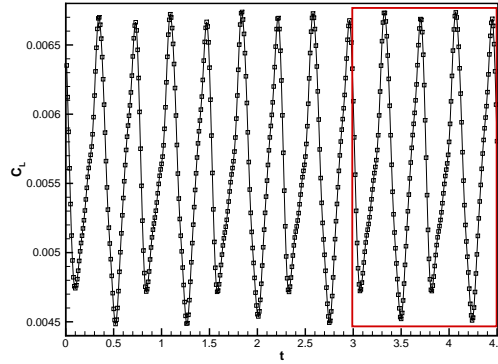
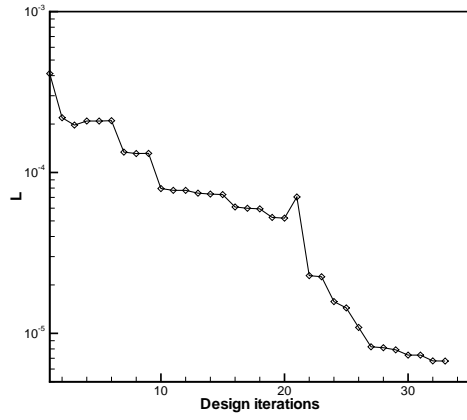


Figure 17. Profile of time-dependent lift coefficient for subsonic flow over the initial NACA0012 airfoil with a $0.03c$ blunt trailing edge, where the red box indicates the objective time-interval.

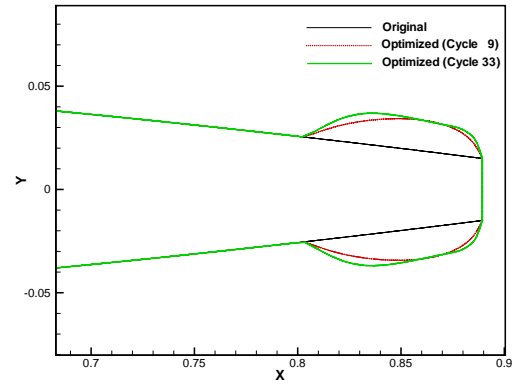
$$\bar{p}_q = \left(\frac{1}{2} p_q^{n_s} + \sum_{n=n_s+1}^{N-1} p_q^n + \frac{1}{2} p_q^N \right) / T^* \quad (42)$$

where T^* denotes the time interval from time step n_s to time step N when pressure fluctuations are measured. In this test case, we first initialize the flow using the uniform flow condition in the far-field and then march the unsteady solution until it becomes periodic. Then we save the periodic solution as the initial flow condition to start the unsteady discrete adjoint-based design optimization. A time-step size of 0.01 and a total of 450 time steps are selected in the example, however, only the last 155 time steps, which correspond to approximately two periods of pressure fluctuations, are related to the time interval in which the pressure noise (i.e. the objective functional) is measured. This is because an adjustment time interval must be considered so that the pressure waves generated by the shedding vortices at the airfoil trailing edge could propagate through the one-chord circular locations specified. It is also noted that little or no shape change may be produced without a reasonable length of time-integration. Figure 17 depicts the time-dependent lift coefficient profile for subsonic flow over the original NACA0012 blunt trailing edge airfoil and the red box in this figure indicates the actual time-interval for calculation of the objective functional.

A fourth-order (i.e. $p = 3$) discontinuous Galerkin scheme and the second-order BDF2 temporal scheme are employed for the respective spatial and temporal discretizations in this case. The Lax-Friedrichs approximate Riemann solver³³ is used for the flux function at interior edges/boundaries and the particular wall boundary treatment discussed



(a) Objective convergence history



(b) Design shapes

Figure 18. (a) Convergence of the objective functional in terms of the number of design iterations for the blunt trailing edge flow. (b) Comparison of optimized shapes at various design cycles with the original airfoil shape.

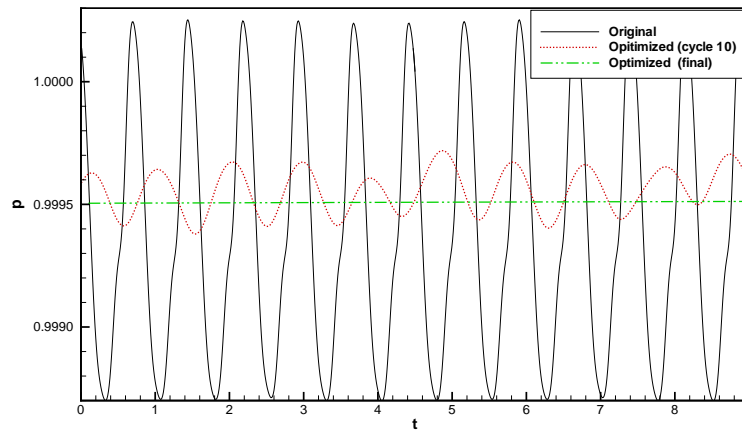
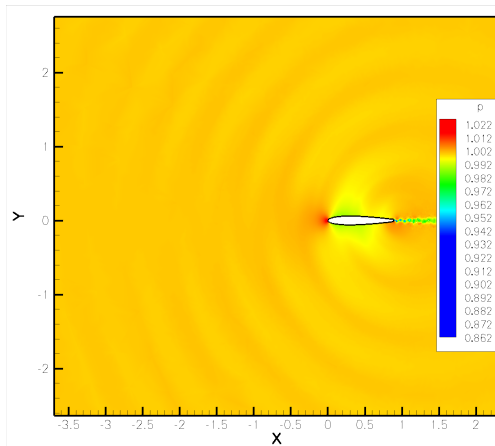
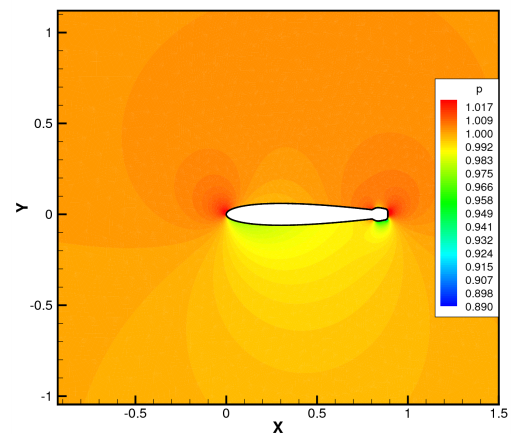


Figure 19. Comparison of pressure signals at a location of one-chord length from the airfoil centerpoint. (The pressure fluctuations produced by the optimized shapes are shifted close to the mean of the original pressure signals for comparison.)



(a) Original airfoil



(b) Final airfoil

Figure 20. Comparison of pressure solutions for the original and final optimized airfoil geometries in the case of airfoil shape design for blunt trailing edge acoustic radiation.

in Section II is employed at wall/surface boundaries. The original symmetric airfoil geometry and the flow with zero-degree angle of attack lead to the generation of shedding vortices from the upper and lower trailing edges with a similar frequency. Thus the present test case follows a consistent shape deformation procedure throughout the design optimization process in order to maintain a symmetric optimized airfoil for simplicity. This is achieved by setting design variables only on the lower airfoil surface, as indicated by the red dots in Fig. 16 (b), and the deformation of the mesh points as well as surface quadrature points on the upper airfoil surface is performed via the Hicks-Henne bump function and values of the design variables obtained from the lower airfoil configuration. In particular, 12 design variables are utilized and surface deformation is only allowed to occur on the upper and lower airfoil surfaces within 10 percent of the chord length from the trailing edge.

The convergence history of the objective functional for the noise minimization design problem is presented in Fig. 18 (a). Due to the fact that very skewed elements tend to be generated particularly in regions near the upper and lower trailing edges if relatively large bounds are selected for the design variables, reasonable bounds must be specified initially to avoid failure of the flow solver or generation of collapsed surfaces. The optimization performance, however, is limited by the bounds since the design variables still require more freedom in movements to further decrease the objective. Therefore, in order to improve the quality of the deformed computational mesh, an edge-swapping procedure and a mesh smoothing technique¹² are employed at the 10th design iteration, which enables the objective convergence to achieve about one order of magnitude reduction by the 20th design iteration. Additionally, while a periodic flow solution for the original airfoil geometry is used as an initial condition in the following design iterations to limit the overall computational effort, this flow solution can not represent an accurate solution for the flow field if there is any change in the computational mesh, hence an updated periodic solution is re-computed every ten design iterations based on the corresponding working mesh and optimized geometry. As seen in Fig. 18 (a), this procedure introduces a spike at the 20th design iteration in the convergence history, but it contributes to a significant improvement in the objective convergence and a two-order magnitude reduction in the objective is finally achieved in approximately 30 design steps. Fig. 18 (b) displays comparison of the optimized airfoil shapes at design cycle 9 and 33 (i.e. final) with the original airfoil geometry. It is observed that the designed shapes attempt to increase the trailing edge angles to eliminate generation of vortices, while still keeping a smooth surface shape. Fig. 19 illustrates the pressure fluctuations produced by the original airfoil and optimized airfoils at a location of $1c$ below the centerpoint of the airfoil, where the pressure disturbance curves for the designed shapes are shifted close to the mean of the original pressure signature for comparison. One can clearly see that the generated noise by the original blunt trailing edge airfoil is reduced by about 70 percent after only 10 design cycles, and furthermore, the optimized shape at the final design cycle produces zero noise or pressure fluctuations at this location since the corresponding flow field finally reaches a steady-state condition. Fig. 20 illustrates the pressure solution contours for the original and final optimized airfoil geometries, where it is clearly shown that a steady-state solution is eventually achieved by the final designed airfoil shape.

V. Conclusions

This work presents a detailed derivation of an unsteady discrete adjoint formulation for high-order discontinuous Galerkin discretizations and investigates areas of unsteady aerodynamic shape optimization by calculating adjoint-based sensitivity gradients to minimize a specified objective functional. The major computational cost at each design iteration for the adjoint-based sensitivity calculation arises from both the flow solution and the flow-adjoint solution, and the rest of the linearization calculation only accounts for about 5 percent of the computational effort. This implies that significant computational efficiency can be in fact gained using the developed adjoint techniques compared to a finite-difference sensitivity calculation where each design variable requires at least one additional flow analysis simulation. On the other hand, the flow-adjoint problem must be solved based on the same temporal scheme which is used in the flow marching scheme, and the definition of the adjoint problem corresponds to a series of linear systems formulated by the transpose of the unsteady Jacobian matrix. Furthermore, the evaluation of sensitivity derivatives requires linearization of geometric mapping coefficients with respect to both mesh points and extra surface quadrature points for high-order curved boundary elements in the discontinuous Galerkin discretizations. The present mesh deformation strategy involves a similar deformation technique for the additional surface quadratures as for the standard surface mesh points. This has been used successively herein for inviscid flow design problems, however, a more sophisticated method may be required for viscous flow design problems where highly stretched and curved elements may be used in boundary layer regions.

In addition, the present work also investigates capabilities of the current implicit DG solver for the aeroacoustics gust response problem where accurate time-dependent flow solutions are required for both aerodynamics and aeroacoustics. Results show that the solutions of pressure disturbances on a symmetric Joukowski airfoil surface as

well as the acoustic intensity at three far-field locations provide good agreement with the results from other validated aeroacoustics codes. Future work will concentrate on the development of three-dimensional high-order accurate DG aeroacoustics codes and on the extension of the current unsteady discrete adjoint strategy to more complex three-dimensional design cases.

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