

Progress on the Development of Efficient Methods for Computational Aeroelasticity and Aerodynamic Design Optimization

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COMPUTATIONAL FLUID DYNAMICS LABORATORY



Acknowledgments

- Many thanks to our sponsors; including the National Science Foundation for its support for our work presented today.
- This work would not have been possible without the hard work and countless hours of research/code-development/debugging of our team over the years.
 - Dr. Reza Djeddi (UTK, Postdoc)
 - Dr. Hang Li (Duke U., Postdoc)
 - Mr. Andrew Kaminsky (CFDRC, finishing PhD, Senior Research Scientist)
 - Dr. Jason Howison (The Citadel, Assoc. Prof.)
 - Dr. Huang Huang (Northwestern Poly. U., Assoc. Prof.)
 - Dr. Emily Buckman (Raytheon Missile Sys, Senior Eng.
 - Dr. Franklin Curtis (Oak Ridge National Lab, Senior Res
 - Mr. Ali Nejad (NCSU-Havelock, finishing PhD, Lecturer
 - Mr. Matthew Whisenant (PhD student)
 - Mr. John Thress (PhD student)
 - Mr. Coleman Floyd (PhD student)





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- 1) Flutter and LCO Predictions Using our State-of-the-Art "One Shot" Method
- 2) Adjoint-Based Sensitivity Analysis Using our Novel FDOT Automatic Differentiation Tool for Aerodynamic Design Optimization
- 3) Machine Learning and Artificial Neural Networks for Unsteady Flow Prediction

4) Conclusion



Computational Aeroelasticity using the One-Shot Method Introduction





Computational Aeroelasticity

The Collapse of Tacoma Narrows Bridge (Nov. 7th, 1940) Transverse mode excitation followed by a torsional mode vibration caused the failure.





Computational Aeroelasticity





Flutter and Limit Cycle Oscillation

> Flutter:

The onset point of self-excited vibration (Stability problem)

Limit Cycle Oscillation (LCO) :

The vibration following the flutter point having a finite amplitude (Response problem)



Benign LCO response

Explosive LCO response

Flutter test of the DG-300 Glider



Aeroelastic Governing Equations

Structure Dynamics:

 $M\ddot{q} + T\dot{q} + Kq = P(q, t)$

Fluid Dynamics: Reynolds-Averaged Navier-Stokes (RANS) equations closed by the one equation Spalart-Allmaras turbulence model

$$\frac{\partial \boldsymbol{U}}{\partial t} + \frac{\partial \boldsymbol{F}}{\partial x} + \frac{\partial \boldsymbol{G}}{\partial y} + \frac{\partial \boldsymbol{H}}{\partial z} = \boldsymbol{S}$$

$$\boldsymbol{U} = \begin{bmatrix} \rho \\ \rho u \\ \rho v \\ \rho w \\ \rho E \\ \rho \tilde{\nu} \end{bmatrix}, \quad \boldsymbol{F} = \begin{bmatrix} \rho u - \rho \dot{f} \\ \rho u^{2} + p - \tau_{xx} - \rho u \dot{f} \\ \rho uv - \tau_{xy} - \rho v \dot{f} \\ \rho uw - \tau_{xz} - \rho w \dot{f} \\ \rho uW - \tau_{xz} - \rho w \dot{f} \\ \rho uW - \tau_{xh} - \rho E \dot{f} \\ \rho u\tilde{\nu} - \tau_{x\nu} - \rho \tilde{\nu} \dot{f} \end{bmatrix}, \quad \boldsymbol{G} = \begin{bmatrix} \rho v - \rho \dot{g} \\ \rho uv - \tau_{yx} - \rho u \dot{g} \\ \rho v v - \tau_{yz} - \rho w \dot{g} \\ \rho v W - \tau_{yz} - \rho w \dot{g} \\ \rho v \tilde{\nu} - \tau_{y\nu} - \rho \tilde{\nu} \dot{g} \end{bmatrix}, \quad \boldsymbol{H} = \begin{bmatrix} \rho w - \rho \dot{h} \\ \rho uw - \tau_{zx} - \rho u \dot{h} \\ \rho w v - \tau_{zy} - \rho v \dot{h} \\ \rho w H - \tau_{zh} - \rho E \dot{h} \\ \rho w \tilde{\nu} - \tau_{z\nu} - \rho \tilde{\nu} \dot{h} \end{bmatrix}, \quad \boldsymbol{S} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ S_t \end{bmatrix}$$



Huang and Ekici (2013), Howison and Ekici (2014)



Harmonic Balance (HB) Method



Hall et al. (2002, 2013), Ekici and Huang (2012)

- Fourier-based mixed time-frequency domain method
- Capacity of modeling strong nonlinear periodic unsteady flows and structural vibrations by incorporating multiple harmonics
- Significant computational cost savings by transferring unsteady problem into mathematically stable problem
- More convenient to analyze the flutter and LCO problems



Harmonic Balance (HB) Method





Structural Dynamics Equations

Original structural dynamics equation:

$$M\ddot{q} + T\dot{q} + Kq = P(q, t)$$

$$\tilde{\omega}^2 D^2 M^* q^* + \tilde{\omega} DT^* q^* + T^* q^* - P^* = 0$$
Unstable when solved in pseudo-time !

State-space formulation:

time derivative:

$$\dot{\eta} + A\eta + Rf = 0$$

$$\eta = \begin{bmatrix} q \\ \dot{q} \end{bmatrix}, \quad A = \begin{bmatrix} 0 & | & -I \\ M^{-1}K & M^{-1}T \end{bmatrix}, \quad R = \begin{bmatrix} 0 & 0 \\ 0 & | & -M^{-1} \end{bmatrix}, \quad f = \begin{bmatrix} 0 \\ P \end{bmatrix}$$
Approximate time derivative by pseudo-spectral operator:

$$\tilde{\omega}D\eta^* + A^*\eta^* + R^*f^* = 0$$
Equation with pseudo-

$$\partial \eta^* \qquad \tilde{\sigma} \eta^* \qquad \tilde{\sigma} \eta^* \qquad \tilde{\sigma} \eta^* \qquad \tilde{\sigma} \eta^* = 0$$

$$rac{\partial oldsymbol{\eta}^*}{\partial au_s}+ ilde{\omega}oldsymbol{D}oldsymbol{\eta}^*+oldsymbol{A}^*oldsymbol{\eta}^*+oldsymbol{R}^*oldsymbol{f}^*=oldsymbol{0}$$



The "One-Shot" Approach

Equation with pseudotime derivative:

$$rac{\partialoldsymbol{\eta}^*}{\partial au_s}+ ilde\omegaoldsymbol{D}oldsymbol{\eta}^*+oldsymbol{A}^*oldsymbol{\eta}^*+oldsymbol{R}^*oldsymbol{f}^*=oldsymbol{0}$$

- The essence of the One-shot approach is to determine the value of the reduced frequency by minimizing the residual of the structural dynamics equation using an optimization.
 - The structural dynamics equation is much simpler (ODE) to deal with
 - It can be solved using a very efficient implicit Euler method with a global "Structural" pseudo timestep allowing values as high as 100.
 - The difficulty due to the nonlinearity (the term *f* due to the generalized aerodynamic forces) in the ODE can be mitigated by lagging the term by one pseudo time iteration.
- The resulting technique is very efficient with computational times that are orders of magnitude smaller than a time-accurate approach.

$$Z(\tilde{\omega}) = \frac{1}{2} |\tilde{\omega} D\eta + A\eta + Rf|^2 \longrightarrow \tilde{\omega}_{\text{new}} = -\frac{[\eta^T A^T + f^T R^T] D\eta}{\eta^T D^T D\eta}$$



Aeroelastic Governing Equations

Fluid Dynamics

$$\frac{\partial \boldsymbol{U}^{*}}{\partial \tau_{f}} + \tilde{\omega} \boldsymbol{D} \boldsymbol{U}^{*} + \frac{\partial \boldsymbol{F}^{*}}{\partial x} + \frac{\partial \boldsymbol{G}^{*}}{\partial y} + \frac{\partial \boldsymbol{H}^{*}}{\partial z} = \boldsymbol{S}^{*}$$

$$\frac{\partial \boldsymbol{U}}{\partial \tau_{f}} + \mathcal{R}_{f} \left(M_{\infty}, Re_{\infty}, \boldsymbol{\eta}(\bar{q}_{r}, \phi_{r}, r \in [1, R]), \tilde{\omega} \right) = \boldsymbol{0}$$
Fixed parameters

Structure Dynamics

$$rac{\partial oldsymbol{\eta}^*}{\partial au_s} + ilde{\omega} oldsymbol{D} oldsymbol{\eta}^* + oldsymbol{A}^* oldsymbol{\eta}^* + oldsymbol{R}^* oldsymbol{\eta}^* + oldsymbol{R}^* oldsymbol{f}^* oldsymbol{H}^* oldsymbol{H}^* = oldsymbol{0}$$
 $rac{\partial oldsymbol{\eta}}{\partial au_s} + \mathcal{R}_s \left(oldsymbol{M}, oldsymbol{T}, oldsymbol{K}, \mu, oldsymbol{f}(oldsymbol{U}), ilde{V}, ilde{\omega}
ight) = oldsymbol{0}$
Fixed parameters

Compact form:
$$\mathcal{F}\left(f, \eta(\bar{q}_r, \phi_r, r \in [1, R]), \tilde{V}, \tilde{\omega}\right) = \mathbf{0}$$

Assume both flow and structural vibrations share the same frequency, i.e. frequency "lock-in"



Flowchart of the One-Shot Method





Computational Aeroelasticity Results

1-DOF VIV 2-DOF Pitch-Plunge Airfoils AGARD 445.6 Wing





One-Shot Aeroelastic Results

1-DOF Vortex Induced Vibration (VIV)

Elastically Supported Circular Cylinder in Two-Dimensional Laminar Cross-Flow



Anagnostopoulos and Bearman (1996)

- > Vortex shedding remains 2D and laminar $Re_{\infty} < 180$
- Strouhal frequency is used

 $St = \tilde{\omega}/2\pi$

> Constant damping and linear elasticity

$$m_{h}\ddot{h} + T_{h}\dot{h} + K_{h}h = q_{\infty}DsC_{l}$$

$$\tilde{h}'' + \zeta_{h}\frac{4\pi}{\kappa Re}\tilde{h}' + \frac{4\pi^{2}}{\kappa^{2}Re_{\infty}^{2}}\tilde{h} = \frac{2}{\pi\mu}C_{l} \quad (\zeta_{h} = 0.00136; \quad \kappa = 0.049772; \quad \mu = 149.0913)$$

Nondimensionalized in terms of *Reynolds number* which functions similarly to the reduced velocity



 257×129



(Carlson et al., 2005; Besem et al., 2016)

One-Shot Aeroelastic Results

1-DOF Vortex Induced Vibration (VIV)

- **Case 1:** $Re_{\infty} = 110$
- Sweep over the Reynolds number and determine the LCO conditions (amplitude/frequency) of the system using the one-shot approach







2-DOF Pitch-Plunge Airfoil





Static equation:

 $K_{\alpha}(\alpha_0 - \alpha_{e0}) = q_{\infty}c^2 s C_{m0}$

Unsteady equations:

$$m_h \ddot{h} + S_\alpha \ddot{\alpha} + T_h \dot{h} + K_h h = -q_\infty csC_l$$
$$S_\alpha \ddot{h} + I_\alpha \ddot{\alpha} + T_\alpha \dot{\alpha} + K_\alpha \alpha = q_\infty c^2 sC_m$$

	LCO, Schewe and Deyhle (1996)	Flutter, Isogai (1979,1981)
airfoil geometry	supercritical NLR 7301	symmetric NACA 64A010
plunging mass-wing mass ratio, m_h/m_0	5.729	1.0
elastic axis position, e	0.25	-0.5
airfoil static unbalance, x_{α}	0.555	1.8
radius of gyration (squared), r_{α}^2	0.822	3.48
natural frequency ratio, ω_h/ω_{α}	1.83	1.0
plunge damping coefficient, ζ_h	0.0175	0.0
pitch damping coefficient, ζ_{α}	0.00411	0.0
mass ratio, μ	172.0	60.0
freestream Mach number, Ma_{∞}	0.75	multiple Mach numbers
flow condition	viscous & turbulent	inviscid



2-DOF NLR 7301 Airfoil



3.5



Computational Efficiency

2-DOF NLR 7301 Airfoil

 \blacktriangleright Stable LCO condition $\tilde{V} = 3.2$



2



Time-accurate, $\Delta t = 0.05$

2-DOF NACA 64A010 Airfoil





 257×129 inviscid mesh





Flutter Prediction for AGARD 445.6 Wing



Mode 1 (first bending) - 9.6 Hz

Mode 2 (first torsion) - 38.1 Hz



Mode 4 (second torsion) - 98.5 Hz



Yates (1963, 1985)

Mode 3 (second bending) - 50.7 Hz

Flutter Prediction for AGARD 445.6 Wing





Flutter Prediction for AGARD 445.6 Wing

Comparison of Inviscid and Viscous Surface Pressure Distributions





Flutter Prediction for AGARD 445.6 Wing

LCO response $M_{\infty} = 0.960, \ \bar{q}_1 = 0.01$





Aerodynamic Design Optimization Introduction





Aerodynamic Design Optimization

Computational Fluid Dynamics (CFD):

- Low-Fidelity: Blade-Element-Momentum
- Moderate-Fidelity: Euler and Navier-Stokes Solvers
- High-Fidelity: (U)RANS Solvers, DES, LES, DNS

Design Optimization:

Non-Gradient-Based

- Evolution Strategies, Genetic Algorithms, Random Search
 - Repeated cost function evaluations





"evolved antenna" for 2006 NASA ST5 spacecraft designed using an evolutionary algorithm



"gradient descent" approach



• Gradient-Based

- Iterative solution of nonlinear programming problems
 - Faster convergence (less design cycles)
 - Sensitivity (gradient) information is required

Aerodynamic Design Optimization



- Grid-Transparent Unstructured Parallel Solver
- Reynolds-Averaged Navier-Stokes (RANS) Eqn's +
- S-A Turbulence and B-C Transition Models
- Steady, Time-Accurate, Time-Periodic (HB)

- Fast and Fully-Automated Discrete Adjoint Sensitivity Analysis
- Operator-Overloading Technique with OOP
- Computationally and Memory Efficient
- Easy Implementation into any Solver



UNPAC Design Optimization Framework



UNPAC-DOF



1) UNPAC Solver

Governing Equations

Reynolds-Averaged Navier-Stokes Equations:

$$\frac{\partial}{\partial t} \int_{\mathcal{V}} \vec{U} \, d\mathcal{V} + \oint_{\partial \mathcal{V}} \left[\vec{F}_c - \vec{F}_v \right] \, dS = \int_{\mathcal{V}} \vec{Q} \, d\mathcal{V}$$

where $\vec{U} = [\rho, \rho \vec{v}, \rho E, \rho \vec{v}]^{T}$ $\vec{F}_{c} = \begin{bmatrix} \rho V \\ \rho u V + p n_{x} \\ \rho v V + p n_{y} \\ \rho w V + p n_{z} \\ \rho H V \\ \rho \vec{v} V \end{bmatrix}$ $\vec{F}_{v} = \begin{bmatrix} 0 \\ \vec{\tau}_{x_{1}} \cdot \vec{n} \\ \vec{\tau}_{x_{2}} \cdot \vec{n} \\ \vec{\tau}_{x_{3}} \cdot \vec{n} \\ \vec{\sigma} \cdot \vec{n} \\ \vec{\tau}_{x_{3}} \cdot \vec{n} \\ \vec{\sigma} \cdot \vec{n} \\ \vec{\tau}_{x_{n}} \cdot \vec{n} \\ \vec{\sigma} \cdot \vec{n} \\ \vec{\tau}_{x_{n}} \cdot \vec{n} \\ \vec{\sigma} \cdot \vec{n} \\ \vec{\tau}_{x_{n}} \cdot \vec{n} \\ \vec{\sigma} \cdot \vec{n}$

$$\frac{\partial \mathcal{V}_i \vec{U_i}}{\partial t} + \vec{R_i} = 0$$

semi-discretized RANS equations













Minimal changes need to be made in order to convert the primal code to the adjoint code:

Pseudo-code: Nozzle Flow Solver (primal code)

```
1 call start_up
2 call mesh
3 call flow_initialization
4
5 do iter = 1,max_iter
6 call one_iteration
7 end do
8
9 call cost_function
```

Pseudo-code: Nozzle Flow Solver (adjoint code)





Implementation

FDOT Toolbox Module:

- Module is included in source codes
- Real variables are replaced by **AReal** type
- Nominal solver is run and the "fully-converged" solution is used to initiate the adjoint solver
- Using a "*checkpointing*" function, the iterative part of the tape is marked
- Iterative variables are flagged
- Adjoint of the cost function is set to unity with all others initialized by zero
- "Tape evaluation" function is called



 Djeddi, R., and Ekici, K.. "FDOT: A Fast, memory-efficient and automated approach for Discrete adjoint sensitivity analysis using the Operator overloading Technique." *Aerospace Science and Technology 91 (2019): 159-174.*



3) UNPAC-DOF

UNPAC-AD

- UNPAC Adjoint Solver:
 - UNPAC solver coupled with the FDOT toolbox







UNPAC-OPT

Optimizer Program:

- Unbounded and Bound Constrained Optimization
- L-BFGS-B Optimizer:

 $\label{eq:min} \begin{array}{l} \min \, f(\vec{x}) \\ \mbox{subject to} \quad \vec{l} \leq \vec{x} \leq \vec{u} \end{array}$

- <u>Objectives:</u>
 Drag (Minimize)
 Lift (Maximize)
 - Lift/Drag (Maximize)

 $B_p^n(s) = C_p^n s^p (1-s)^{n-p}$

Shape Deformation/Parameterization

- Surface Grid Points:
 - <u>Smoothing</u> process applied to surface perturbations
- Free-Form Deformation (FFD) Box:

$$\mathbf{x}_{s} = \sum_{i=0}^{\text{NI}} \sum_{j=0}^{\text{NJ}} \sum_{k=0}^{\text{NK}} B_{i}^{\text{NI}}(\xi_{s}) \ B_{j}^{\text{NJ}}(\eta_{s}) \ B_{k}^{\text{NK}}(\zeta_{s}) \ \mathbf{x}_{\text{cp}}$$

$$\Delta x_i = x_i^{n+1} - x_i^n$$
S
$$\Delta x_i^* + \sum_{j=1}^{\operatorname{Ngb}_i} \epsilon \left[\Delta x_i^* - \Delta x_j\right] = \Delta x_i$$

$$\vec{x}^{n+1} = \vec{x}^n + \vec{\Delta x}^*$$



3) UNPAC-DOF

UNPAC-DOF

- Wrapper Program:
 - Couples the UNPAC, UNPAC-AD, and UNPAC-OPT programs
 - Written in Modern Fortran
 - Uses "bash scripting" to organize solution files and folders at each design cycle
 - Applies a pseudo-Laplacian to smooth surface perturbations





Aerodynamic Design Optimization Results

NACA0012 Airfoil ONERA M6 Wing





NACA0012 Airfoil Subject to Inviscid Transonic Flow

Drag Minimization:

- M = 0.8, AoA = 1.25 deg Ο
- "Surface Points" used as DV's for deformation Ο





12

1.4

1.3 1.2

1.1 1 0.9

0.8 0.7 0.6

NACA0012 Airfoil Subject to Inviscid Transonic Flow

Drag Minimization:

- M = 0.8, AoA = 1.25 deg
- "2D FFD Box" used for shape parameterization





NACA0012 Airfoil Subject to Inviscid Transonic Flow

Drag Minimization:

- M = 0.8, AoA = 1.25 deg
- Comparison of Shape Parameterization Results





NACA0012 Airfoil Subject to Inviscid Transonic Flow

Drag Minimization:

- M = 0.8, AoA = 1.25 deg
- Comparison of Shape Parameterization Results



Geometry	C_D	Reduction	$t_{\rm max}/c$	Reduction
Original	2.1638E-2	-	0.120	-
Optimized (Surface Points)	9.1755E-4	95.7%	0.093	22.5%
Optimized (FFD Box)	8.7265E-4	95.9%	0.115	4.2%

Memory Footprint:

Adjoint Solver -> <u>800</u> Mbytes



Normalized CPU Time:

ONERA M6 Wing

- Drag Minimization:
 - Based on ONERA D airfoil section (10% t/c)
 - \circ 30 degree sweep angle
 - Aspect Ratio = 3.8, Taper Ratio = 0.562
 - M = 0.8395, AoA = 3.06 deg



NASA TMR: Schmitt and Charpin (1979)





ONERA M6 Wing

Drag Minimization:

• M = 0.8395, AoA = 3.06 deg





ONERA M6 Wing

Drag Minimization:

• M = 0.8395, AoA = 3.06 deg





ONERA M6 Wing



• M = 0.8395, AoA = 3.06 deg



Geometry	C_D	Reduction
Original	1.1734E-2	-
Optimized	8.4601E-3	27.9%

Memory Footprint:

Adjoint Solver -> <u>14.1</u> Gbytes



Normalized CPU Time:

Machine Learning and Neural Networks For Unsteady Flow Predictions





Machine Learning (ML) Artificial Neural Networks (ANN)

Artificial Intelligence Based on ML/ANN



Machine Learning (ML) Artificial Neural Networks (ANN)

Artificial Intelligence Based on ML/ANN

Problem Formulation:

Governing Eqn's:

$$\frac{\partial\omega}{\partial t} + J(\frac{\partial\psi}{\partial\eta}\frac{\partial\omega}{\partial\xi} - \frac{\partial\psi}{\partial\xi}\frac{\partial\omega}{\partial\eta}) = \frac{1}{R_e}(\alpha\frac{\partial^2\omega}{\partial\xi^2} + 2\gamma\frac{\partial^2\omega}{\partial\xi\partial\eta} + \beta\frac{\partial^2\omega}{\partial\eta^2} + P\frac{\partial\omega}{\partial\xi} + Q\frac{\partial\omega}{\partial\eta})$$
$$\alpha\frac{\partial^2\psi}{\partial\xi^2} + 2\gamma\frac{\partial^2\psi}{\partial\xi\partial\eta} + \beta\frac{\partial^2\psi}{\partial\eta^2} + P\frac{\partial\psi}{\partial\xi} + Q\frac{\partial\psi}{\partial\eta} = -\omega$$

Rotationally oscillating cylinder in cross-flow:





Machine Leanner, Artificial Neural Networks (1) Based on ML/ANN

Flow Prediction Results:



"Lock-in case"

"Non-Lock-in case"



Lock-In and Non-Lock-In Regions **Predicted Using ML**





Conclusion

Summary Future Works





Summary: Computational Aeroelasticity

A novel dynamic aeroelastic solution approach, called the One-shot method, is developed, which features the following advantages:

$\ensuremath{\boxdot}$ High efficiency and robustness

Both flow and structure solvers converge simultaneously, and numerical convergence is ensured for a wide range of initial guesses

☑ DOF-independent computational cost

☑ Free from time-synchronization

Both flow and structure solvers preserve relative independence while being coupled in pseudo-time, which greatly improves the efficiency

☑ Flexible inputs

Prescribe either aerodynamic parameter (velocity) or structure parameter (amplitude) as the input; Capacity of resolving both benign and detrimental types of LCO, and flutter onset point using one solver



Summary: Computational Aeroelasticity

☑ Frequency (and Velocity) search procedure

Determine self-excited flutter and LCO;

☑ Implicit CSD solver

Significantly reduces the artificial energy at interface of CFD and CSD solvers, and stabilizes the coupled aeroelastic solver;

☑ Easy implementation

Partitioned code-coupling can be done based on off-the-shelf standalone solvers; No dynamic mesh technique is needed, and no Jacobian terms are involved;

The One-shot method has been applied to model various aeroelastic systems. Numerical results show that this new approach can rapidly predict flutter boundary and different types of LCO responses, offering a promising tool to solve dynamic aeroelasticity problems



Future Works: Computational Aeroelasticity





Summary: Aerodynamic Design Optimization



 Grid-transparent unstructured 2D and 3D compressible RANS solver UNstructured PArallel Compressible (UNPAC)



- Fast automatic Differentiation using Operator-overloading Technique (FDOT)
- Efficient and fully-automated; requiring minimal changes to the primal solver
- Memory efficient (manageable memory footprint for 2D and 3D cases)
- Computationally efficient (Adjoint/Primal normalized CPU time ~ 2 5x

UNPAC Design Optimization Framework



- Design Optimization Framework UNPAC-DOF Wrapper Program
 - Performs Gradient-Based Design Optimization
 - Uses L-BFGS-B Optimizer Program
 - Unbounded and Bound Constrained Optimization
 - Surface Points (2D) & FFD Box for Shape Parameterization/Deformation



echnia

Future Works: Aerodynamic Design Optimization



- Improving the Memory and Computational Efficiency:
 - Optimizing the tape recording process for the expression tree
 - Use of recursive patterns for reducing tape size (will require more changes to be made to the original solver)
 - using directive functions to automize tape recording without requiring a significant amount of user intervention

One-Shot Method for simultaneous solution of "Primal", "Adjoint", and "Design" problems





Thank You!

Q & A

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Backup Slides



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RAE "A" Wing-Body Configuration





1-DOF Vortex Induced Vibration (VIV)

Sweep over *Reynolds number*



0.8

0.7

x x Anagnostopoulos and Bearman, Exp. (1973)

Besem et al., HB/LCO (2016)



RAE "A" Wing-Body Configuration (AGARD-AR-138, Case 6)

M = 0.9, AoA = 1.0 deg

RAE "A" Wing:

- A.R. = 5.5
- \circ L.E. Sweep Angle = 36.7 deg
- \circ T.E. Sweep Angle = 22.3 deg
- \circ Taper ratio = 0.375
- RAE 101 symmetrical airfoils (untwisted)

Unstructured Grid, ~460K tetrahedra, exte







RAE "A" Wing-Body Configuration (AGARD-AR-138, Case 6)

Mach number and Pressure Contours





RAE "A" Wing-Body Configuration (AGARD-AR-138, Case 6)

Pressure Coefficients at Various Spanwise Location Along the Wing





RAE "A" Wing-Body Configuration (AGARD-AR-138, Case 6)

Pressure Coefficients Distributions for 2 Longitudinal Sections along the body





ROM-Based Convergence Acceleration





ROM-Based Convergence Acceleration

 $\vec{U}^{n+1} = \vec{U}^n + \vec{R}(\vec{U})$

ROM-Based Convergence Acceleration

Model Reduction:

Undata formula:

 \triangleright

$$\frac{d\vec{U}}{dt} + \mathcal{N}(\vec{U}) = 0 \quad \text{where} \quad \Phi^T \frac{d}{dt} \left(\Phi \vec{\xi} \right) + \Phi^T \mathcal{N} \left(\Phi \vec{\xi} \right) = 0 \quad \text{where} \quad \vec{U} \cong \Phi \vec{\xi}$$

OProjection

Correlation-Based Approach

 $\vec{R}(\vec{U}) \approx \mathbf{A}\vec{U} - \vec{b}$

Assumption

$$\boldsymbol{\Phi} = \begin{bmatrix} | & | & | \\ \vec{U}_1 & \vec{U}_2 & \dots & \vec{U}_M \\ | & | & | \end{bmatrix}_{N \times M}$$

$$ec{U}_{ ext{projected}} = \sum_{i=1}^M ec{U}_i \xi_i = \mathbf{\Phi} \, oldsymbol{\xi}$$

Covariance-Based Approach

$$\hat{\Phi} = \begin{bmatrix} | & | & | \\ \vec{U}_1 - \vec{U} & \vec{U}_2 - \vec{U} & \dots & \vec{U}_M - \vec{U} \\ | & | & | \end{bmatrix}_{N \times M} \vec{U}_{\text{projected}} = \hat{\Phi} \vec{\xi} + \vec{U} \qquad \quad \vec{U} = \frac{1}{M} \sum_{i=1}^M \vec{U}_i$$





.2 Norm

 $\vec{R}(\vec{U}_{\text{projection}}) \rightarrow 0$

ROM-Based Convergence

Acceleration Inviscid Transonic Flow Past NACA0012 Airfoil

- ROM-Based Convergence Acceleration:
 - Covariance-based approach
 - 10, 20, 40, and 80 solution snapshots used for projections (over a span of 2000 iterations)
 - Process lagged by 1000 iterations (2 orders of magnitude drop in residual)



^a 10 snapshots every 200 iterations during a cycle of 2,000 itrs. and lagged by 1,000 itrs.







r-Adaptive Mesh Redistribution (AMR)





Inviscid Transonic Flow Past NACA0012 Airfoil

- r-Adaptive Mesh Redistribution (AMR):
 - Convergence and Errors











Case Settings

 $\alpha = 1.25 \deg$

 $M_{\infty} = 0.8$



Lift and Drag Predictions

Grid	C_L	Error	Error (relative)	C_D	Error	Error (rel.)
Yano & Darmofal [217]	0.35169	-	+	0.02262		-
Fully-Refined	0.35096	0.20%	. Contraction	0.02349	3.84%	
Baseline	0.34820	0.99%	1.64%	0.02445	8.09%	4.08%
r-adapted (AMR)	0.35024	0.41%	0.20%	0.02302	1.76%	2.00%

Computational Cost

Grid	CPU Time (s)	Normalized CPU Time
Baseline	215.58	1.00
Fully-Refined	2037.60	9.45
r-adapted (AMR)	306.27	1.42



AGARD-702 CT5 Case

- Inviscid Periodic Flow:
 - Unstructured Grid: Nodes = 5,233
 Cells = 10,216
 - Harmonic Balance Method





Computational Fluid Dynamics Lab, MABE Department



Case Settings

 $\alpha_0 = 0.016 \, \deg$

 $M_{\infty} = 0.755$

AGARD-702 CT5 Case



- Baseline grid: Nodes = 1,837
 - Colle = 2 E 42
 - Cells = 3,542
- Fully-refined grid: Nodes = 5,233
 Cells = 10,216

Baseline Grid



Case Settings

 $M_{\infty} = 0.755$ $\alpha_0 = 0.016 \text{ deg}$ $\alpha_p = 2.51 \text{ deg}$ k = 0.0814

r-Adapted Grid at various sub-time levels

Fully-refined Grid











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AGARD-702 CT5 Case

- r-Adaptive Mesh Redistribution (AMR):
 - Significant improvements in numerical accuracy at a fraction of the computational cost

L2 Norm of Errors in C_{L} and C_{M}











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Computational Cost Comparison

Grid	CPU Time (s)	Normalized CPU Time
Baseline	1,749.1	1.00
Fully-Refined	12,671.2	7.24
r-adapted (AMR)	2,207.2	1.26



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