

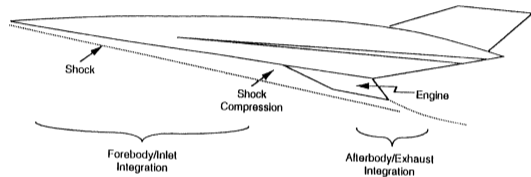
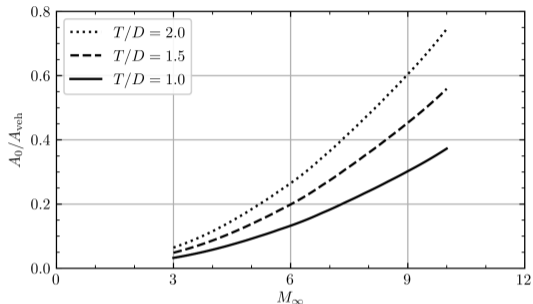
Hypersonic Aerodynamic Shape Optimisation

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Airframe Integration for Hypersonic Vehicles

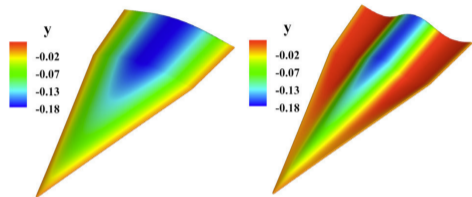


Airframe integration (Schweikart [2]).

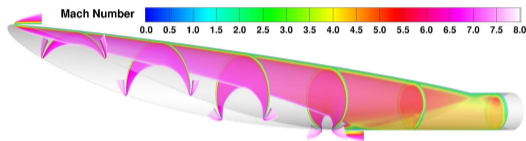
Required capture area vs Mach number (Ward [1]).

- required engine size scales with M_∞
- use airframe to assist with compression & expansion
- components are coupled, geometrically complex, require high-fidelity modelling
 - numerical optimisation

Aerodynamic Shape Optimisation



Optimised hypersonic lifting body (Zhang et al.).



Optimised hypersonic inlet (Drayna).

- two levels of design optimisation: MDO & ASO
- theme in ASO literature: geometric freedom sacrificed for efficiency
- gradient calculation cost for N DVs:
 - finite differences: $N + 1$ flow solutions
 - adjoint method: 1 flow solution + 1 linear system solution
- limited application of adjoint method for 3D hypersonic design

Outline

Thesis aim: *Investigate the applicability of adjoint-based shape optimisation for three-dimensional hypersonic vehicle design.*

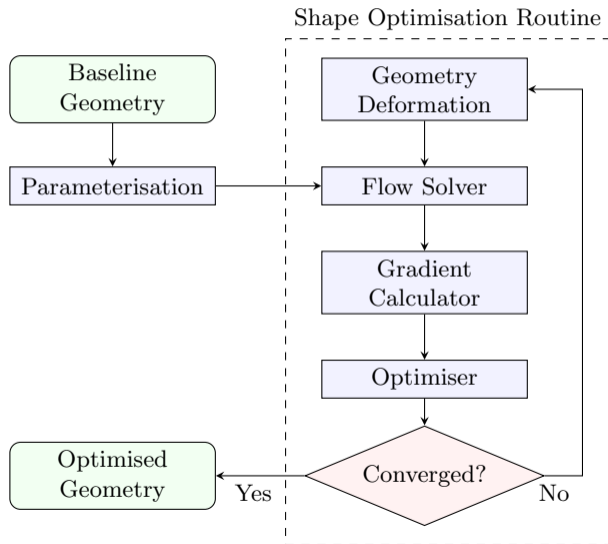
Part 1: Methodology

- geometric manipulation
- gradient calculation

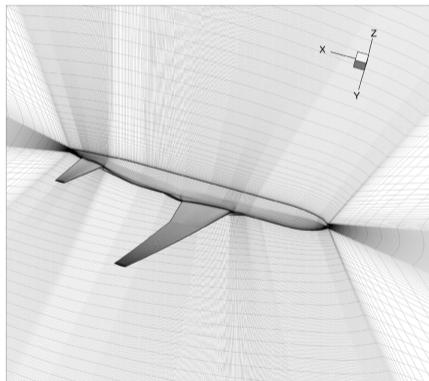
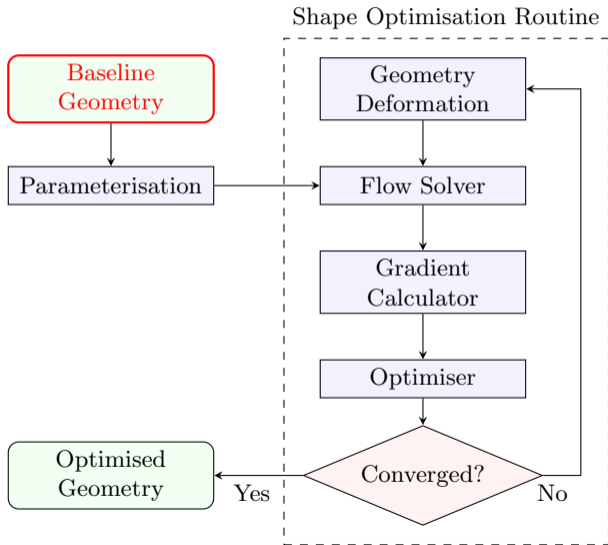
Part 2: Design Applications

- minimum-drag hypersonic slender body of revolution
- hypersonic lifting body

Aerodynamic Shape Optimisation Routine

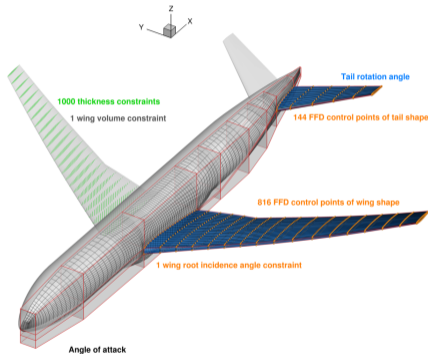
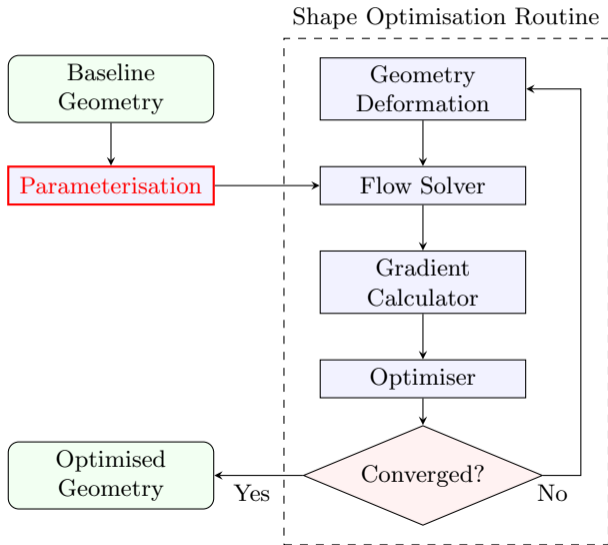


Aerodynamic Shape Optimisation Routine



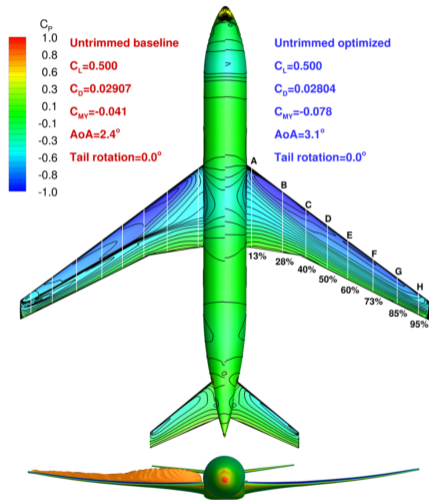
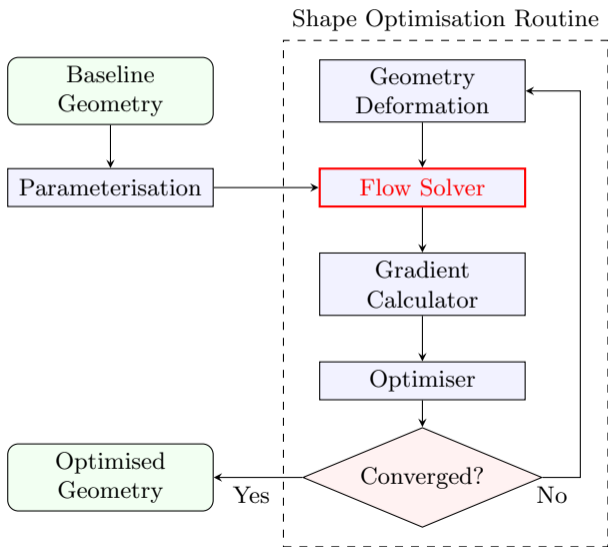
(Chen et al. [5])

Aerodynamic Shape Optimisation Routine



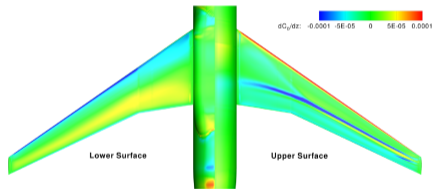
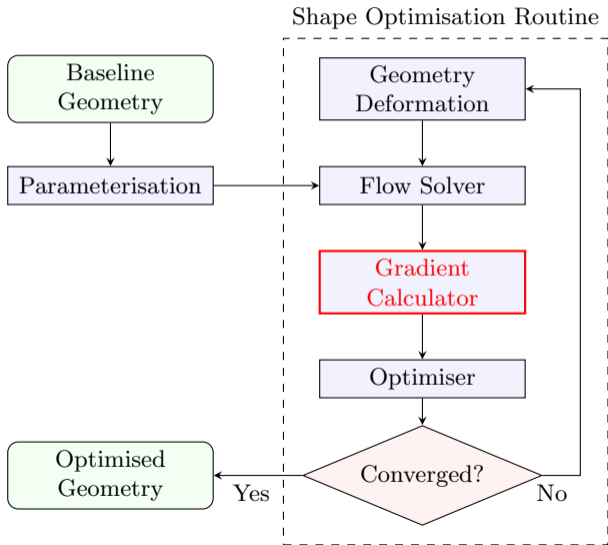
(Chen et al. [5])

Aerodynamic Shape Optimisation Routine



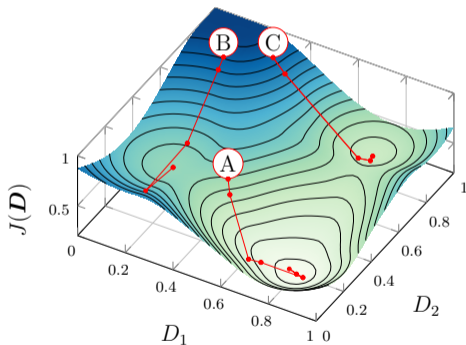
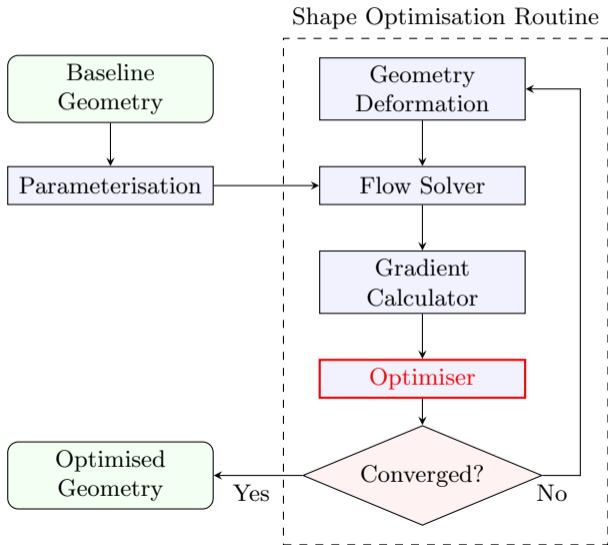
(Chen et al. [5])

Aerodynamic Shape Optimisation Routine

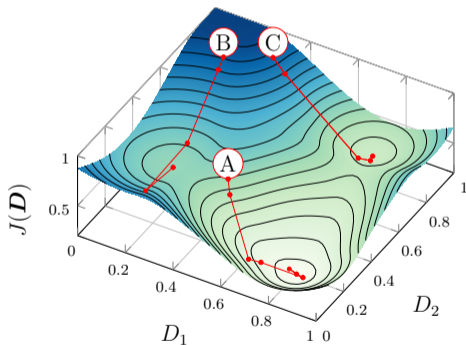
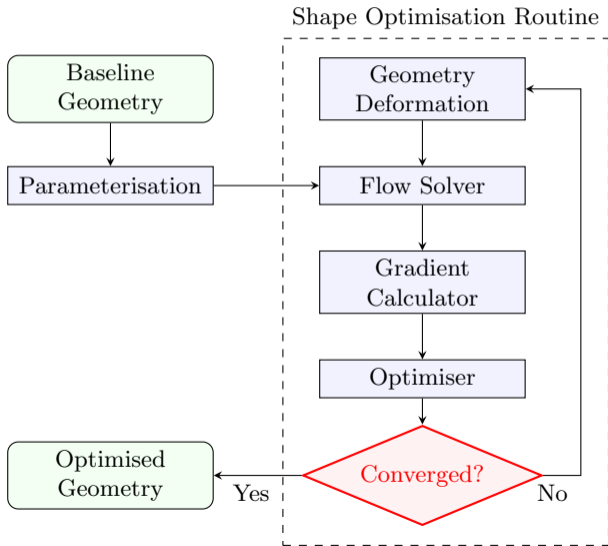


(Chen et al. [5])

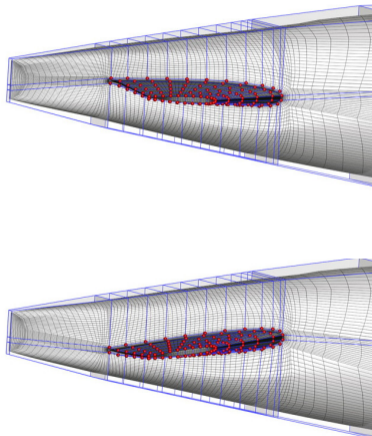
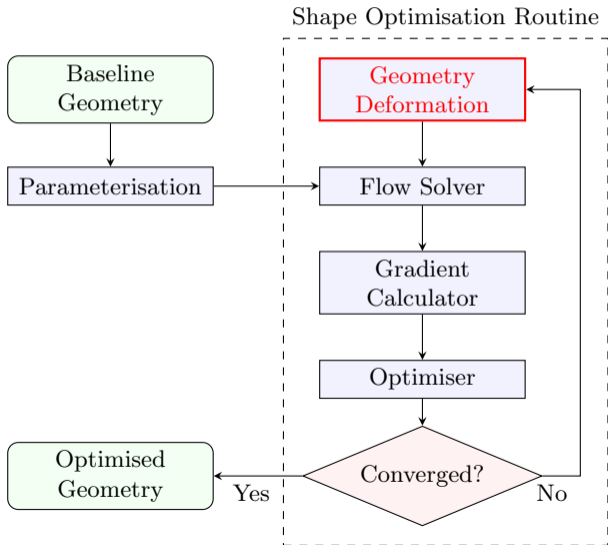
Aerodynamic Shape Optimisation Routine



Aerodynamic Shape Optimisation Routine

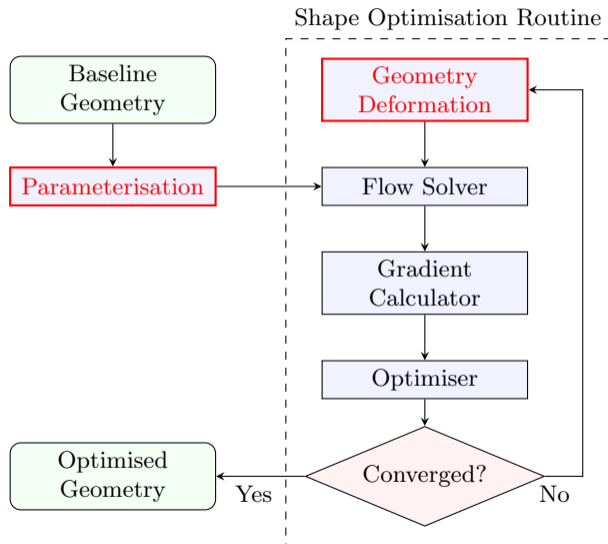


Aerodynamic Shape Optimisation Routine

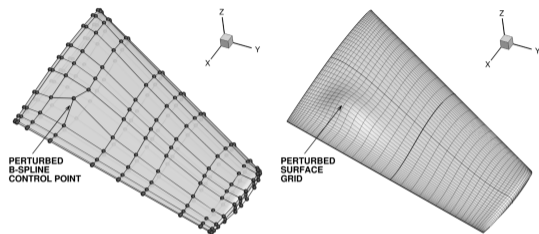


(Chen et al. [5])

Geometric Parameterisation and Deformation

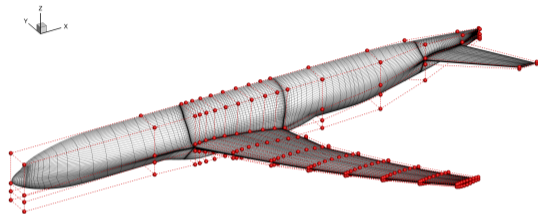


3D Shape Parameterisation Methods



B-Spline surfaces [6]

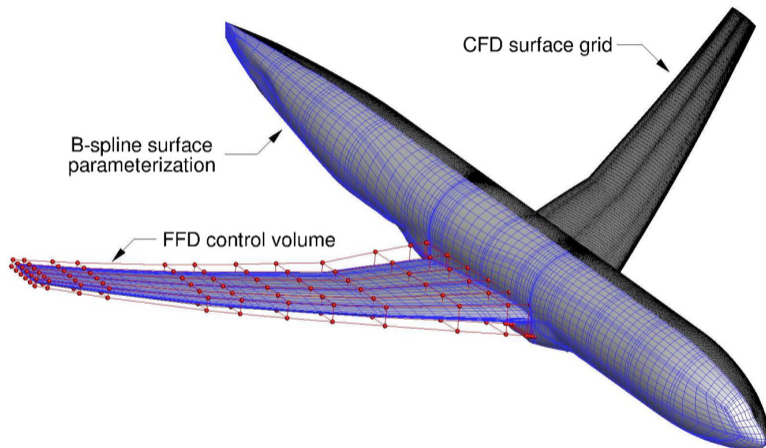
- high local control, analytical representation, allows integrated mesh movement
- susceptible to overlap, tangency issues



Free-form deformation (FFD) [7]

- highly general, robust, DV selection decoupled from surface representation
- discrete representation, separate mesh deformation

Two-level Free-form Deformation



Wing parameterised with two-level FFD (Reist et al. [8]).

Integrated Shape and Grid Deformation

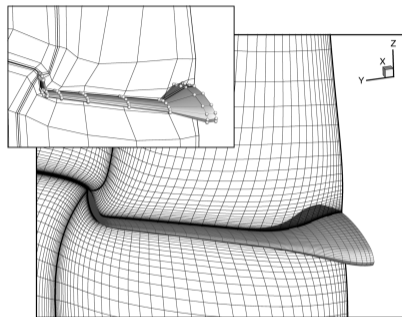
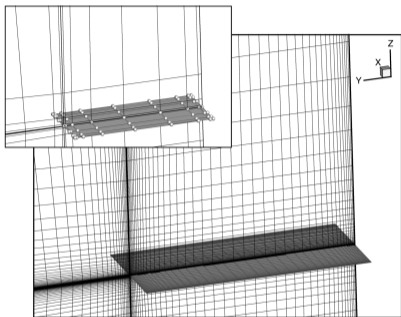
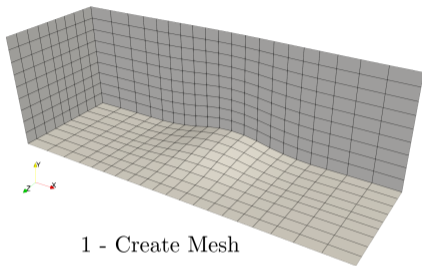


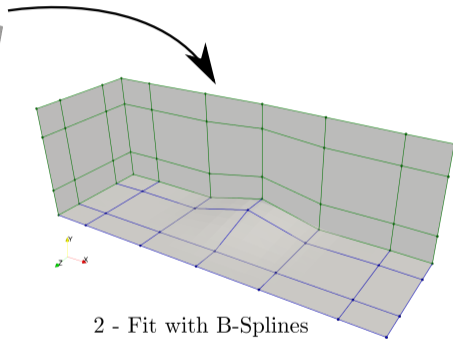
Plate geometry morphed into a blended-wing body (Hicken and Zingg [9]).

- fit entire mesh with B-Splines
- move mesh control points rather than vertices
- use cheap distance-based routine for control points

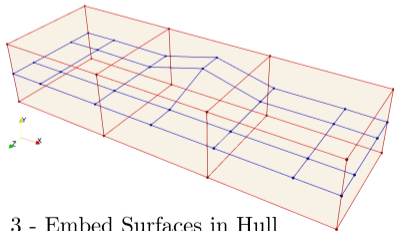
Parameterisation Methodology



1 - Create Mesh



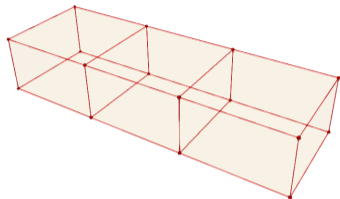
2 - Fit with B-Splines



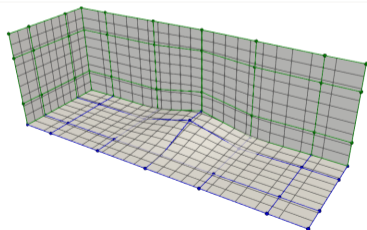
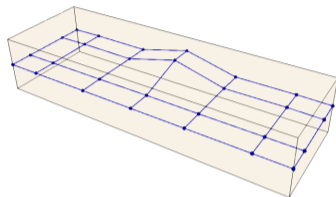
3 - Embed Surfaces in Hull

Deformation Methodology

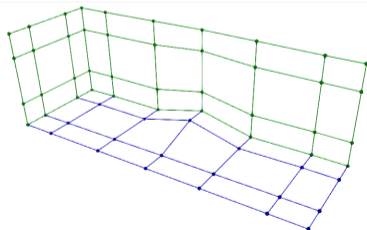
1 - Deform FFD Hull



2 - Update Surface B-Splines

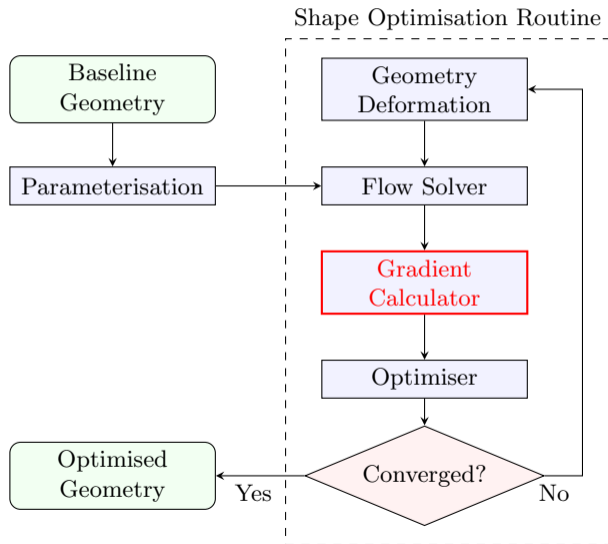


4 - Update Mesh



3 - Update Mesh B-Splines

Gradient Calculation



Shape Sensitivity Calculation

objective function: $J = J(\mathbf{D})$, gradients/shape sensitivities: $dJ/d\mathbf{D}$

Finite differences

$$\frac{\partial J}{\partial D_j} = \frac{J(D_j + h) - J(D_j)}{h} \quad (1)$$

- for each design variable D_j :
 1. perturb variable: $D_j + h$
 2. update surfaces and grid: \mathbf{X}
 3. run flow solver: \mathbf{U}
 4. evaluate function: $J(D_j + h)$
 5. evaluate eqn 1

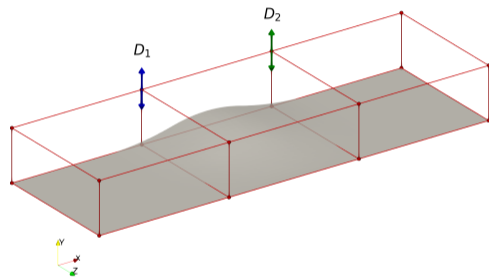
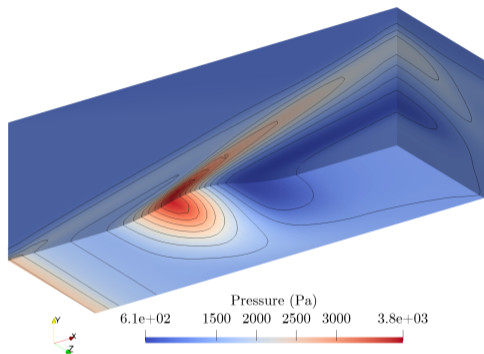
Adjoint method

$$\frac{dJ}{d\mathbf{D}} = \frac{\partial J}{\partial \mathbf{D}} + \boldsymbol{\lambda}^T \frac{\partial \mathbf{R}}{\partial \mathbf{D}} \quad (2)$$

$$\left[\frac{\partial \mathbf{R}}{\partial \mathbf{U}} \right]^T \boldsymbol{\lambda} = - \left[\frac{\partial J}{\partial \mathbf{U}} \right]^T \quad (3)$$

1. compute partial derivatives
2. solve adjoint equations (equation 3)
 - sparse linear system
3. evaluate equation 2

Shape Sensitivity Verification – Hypersonic Bump

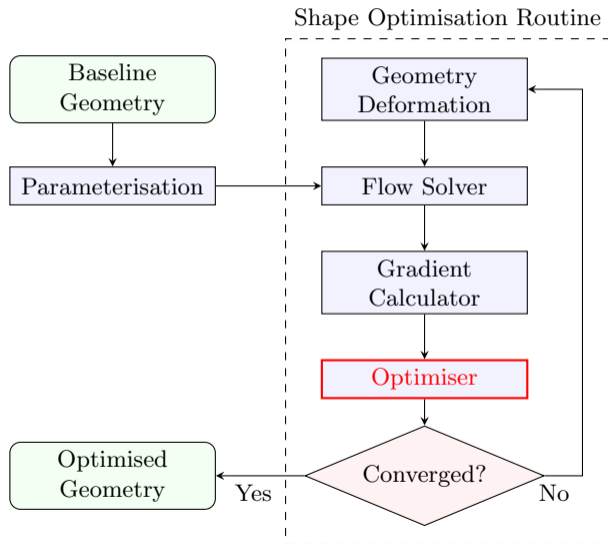


- modelling: Mach 5 flow - Euler, NS & RANS
- objective: $J = F_{D, \text{wave}}$
- design variables: D_1 – vertical translation, D_2 – vertical scaling

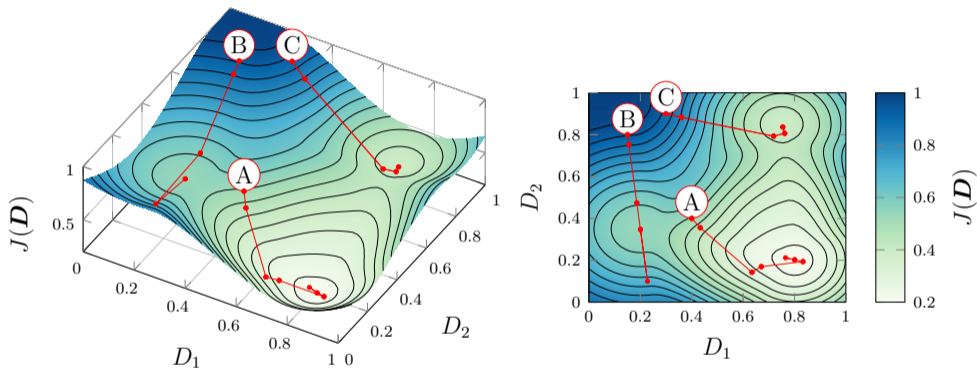
NS Shape Sensitivity Verification

Blocking	Method	$\partial J/\partial D_1$	$\partial J/\partial D_2$
1B	adjoint	7.652439748627 438 e-01	9.8008643528069 10 e-04
	complex-step	7.652439748627 382 e-01	9.8008643528069 64 e-04
2B	adjoint	7.652449114021 621 e-01	9.8008784557861 87 e-04
	complex-step	7.652449114021 596 e-01	9.8008784557861 31 e-04
4B	adjoint	7.6524312915124 93 e-01	9.8008553495246 67 e-04
	complex-step	7.6524312915124 51 e-01	9.8008553495246 28 e-04
8B	adjoint	7.652449084951 116 e-01	9.8008784110591 61 e-04
	complex-step	7.652449084951 1075 e-01	9.8008784110591 22 e-04
27B	adjoint	7.6524292886105 84 e-01	9.8008527369557 65 e-04
	complex-step	7.6524292886105 14 e-01	9.8008527369557 54 e-04

Optimisation Algorithm

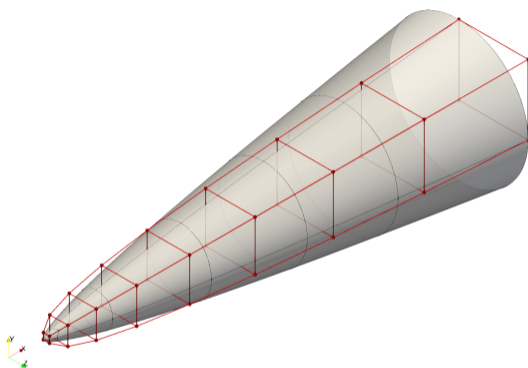
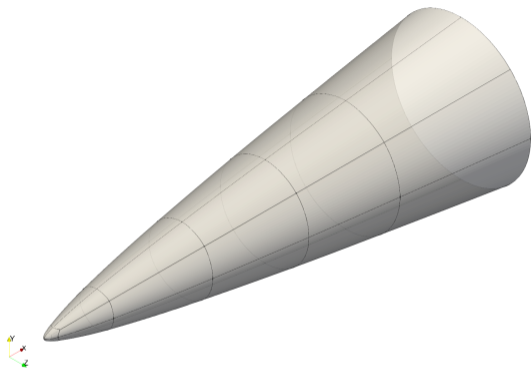


Sparse Nonlinear OPTimizer (SNOPT)



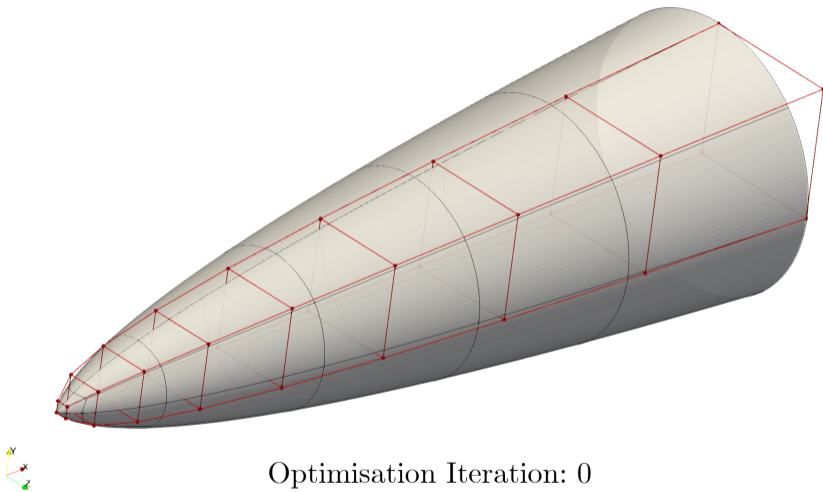
- sequential quadratic programming (SQP) – gradient-based
- constrained non-linear problems
- designed for hundreds to thousands of DVs [10]

Minimum-drag Hypersonic Slender Bodies of Revolution



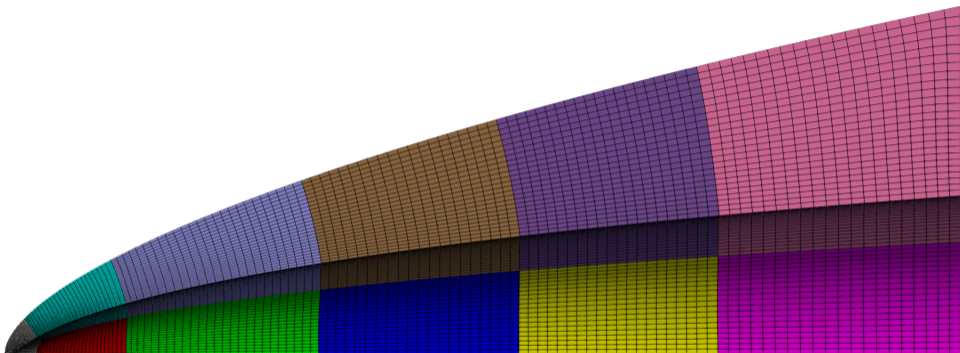
- objective: minimise wave drag
- constraints: fixed length and base diameter ($L/D = 3$)
- parameterisation: 23 DVs – scale and translation of control planes
- modelling: inviscid Mach 6.28 flow

Slender Body Optimisation – Surface Animation



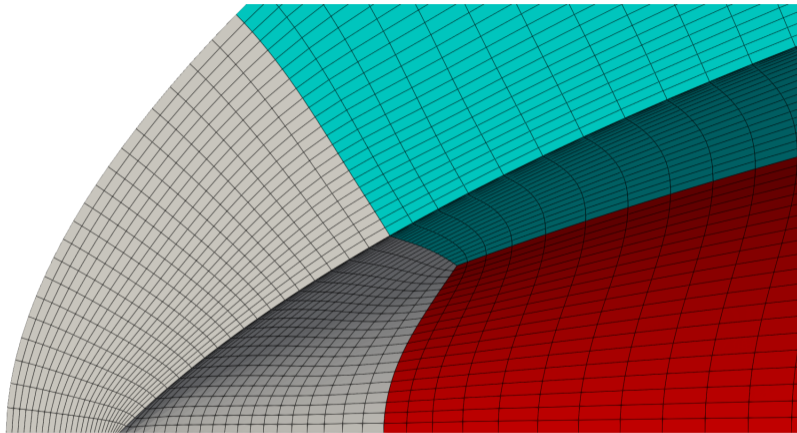
Optimisation Iteration: 0

Slender Body Optimisation – Grid Animation



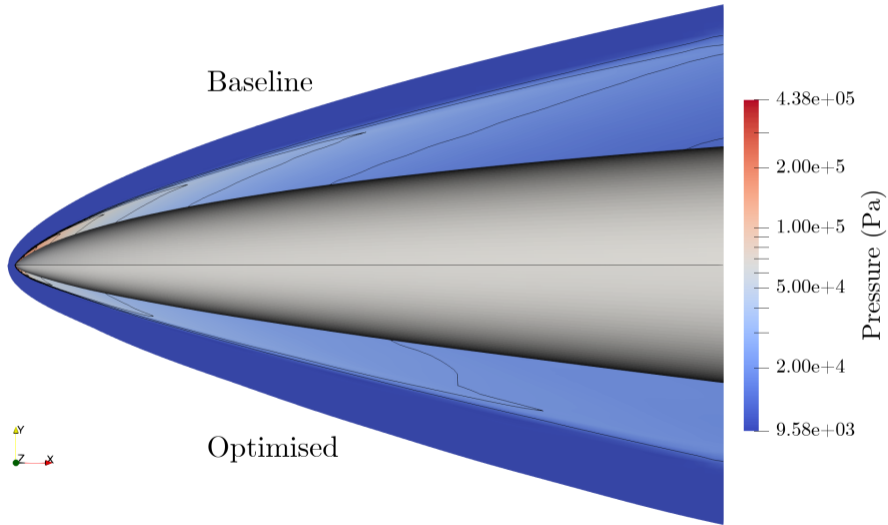
Optimisation Iteration: 0

Slender Body Optimisation – Nose Grid Animation

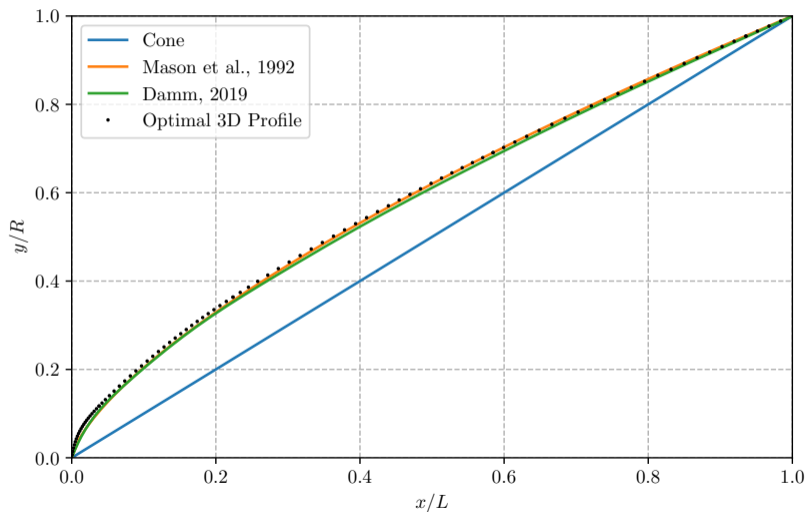


Optimisation Iteration: 0

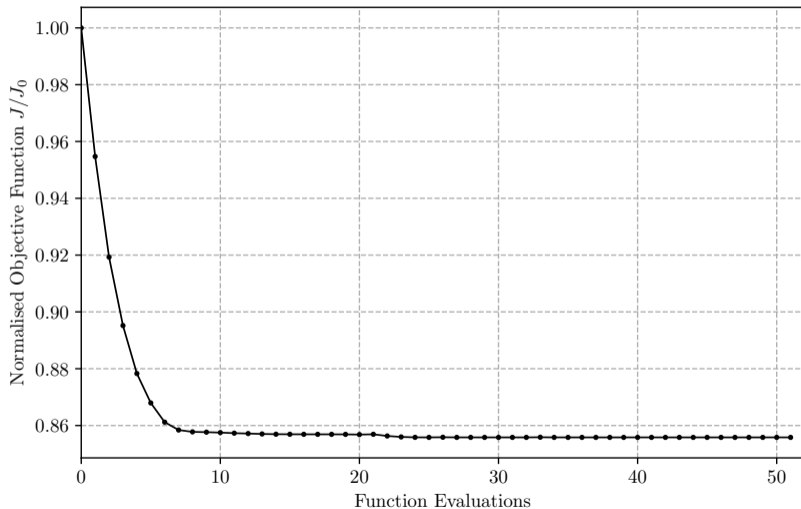
Slender Body Optimisation – Flow Field Comparison



Slender Body Optimisation – Profile Comparison



Slender Body Optimisation – Convergence



Slender Body Optimisation – Breakdown of Costs

Hardware:

- CPU: Intel(R) Xeon(R) Silver 4216 CPU @ 2.10GHz 16 cores
- RAM: 12 × 16GB (192GB) DDR4 2933MHz

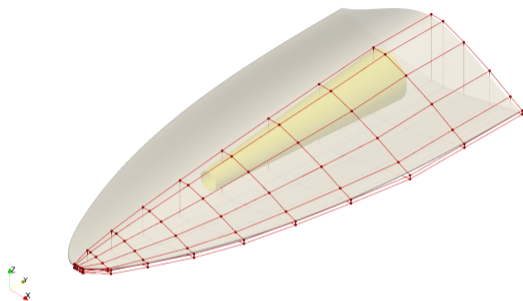
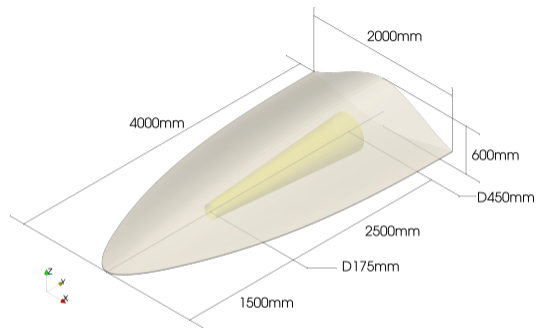
Overall:

- parameterisation: 15s (serial)
- function & gradient evaluations: 52
- total time: 9hrs 40min

Each design iteration:

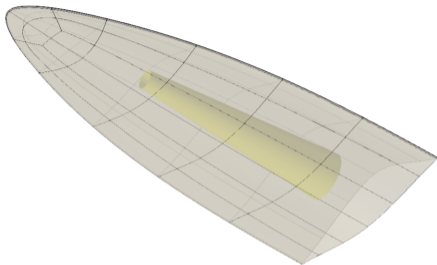
- flow solver: 3min 39s (MPI parallel)
- gradient calculation: 7min 34s (shared-memory parallel)
- geometry deformation: 7s (shared-memory parallel)

Optimisation of a Hypersonic Lifting Body

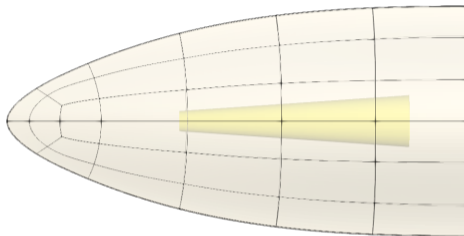


- modelling: $M_\infty = 8$, $\alpha = 8^\circ$, 40km altitude, inviscid, laminar & turbulent
- objective: maximise $J = L/D$
- constraints: fixed length & must contain payload
- parameterisation: 118 FFD DVs & 3 payload DVs

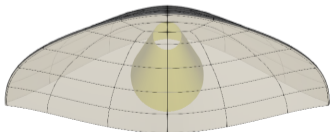
Lifting Body Optimisation – Surface Animation (1/2)



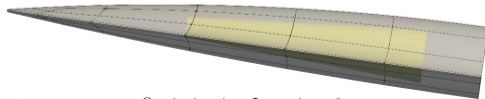
Optimisation Iteration: 0



Optimisation Iteration: 0

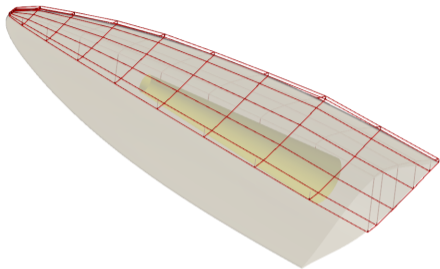


Optimisation Iteration: 0

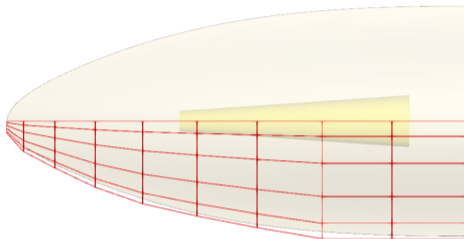


Optimisation Iteration: 0

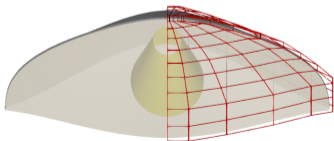
Lifting Body Optimisation – Surface Animation (2/2)



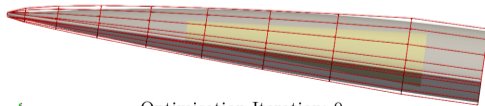
Optimisation Iteration: 0



Optimisation Iteration: 0

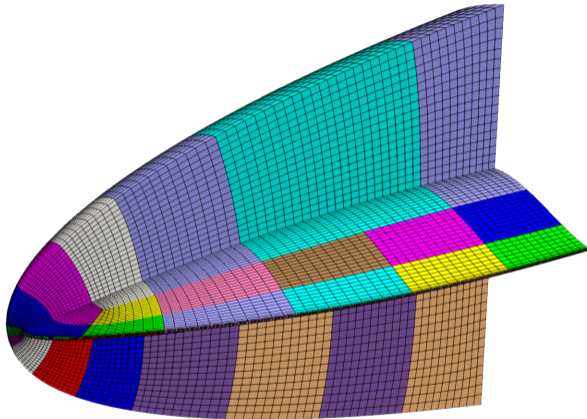


Optimisation Iteration: 0

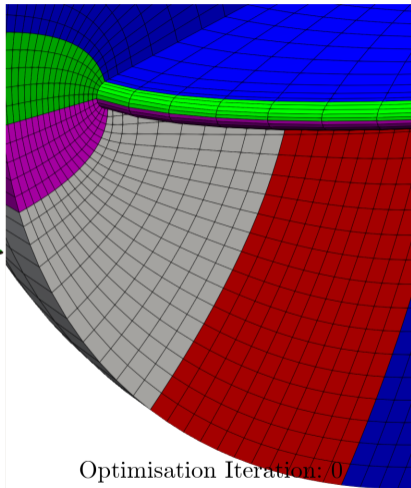


Optimisation Iteration: 0

Lifting Body Optimisation – Grid Animation



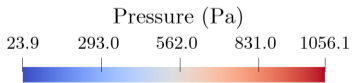
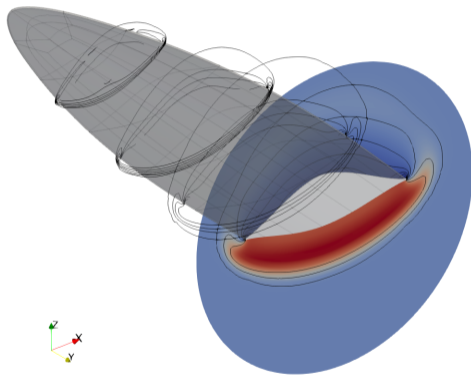
Optimisation Iteration: 0



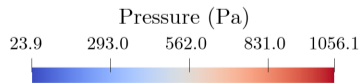
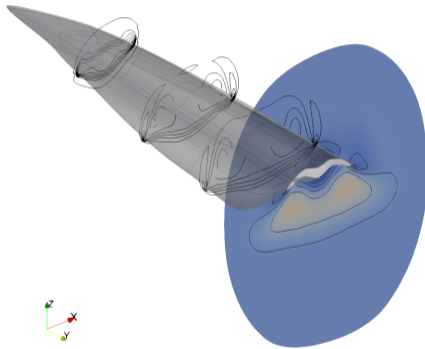
Optimisation Iteration: 0

Lifting Body Optimisation – Comparison to Baseline

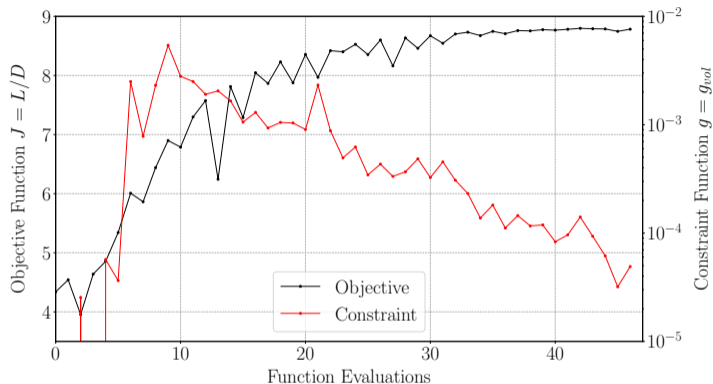
Baseline



Optimised



Lifting Body Optimisation – Convergence



Configuration	Lift L (N)	Drag D (N)	Lift-on-drag L/D
Baseline	9647	2169	4.447
Optimised	2254	258.5	8.719

Lifting Body Optimisation – Breakdown of Costs

Hardware:

- CPU: Intel(R) Xeon(R) Silver 4216 CPU @ 2.10GHz 16 cores
- RAM: 12 × 16GB (192GB) DDR4 2933MHz

Overall:

- parameterisation: 11s (serial)
- function & gradient evaluations: 47
- total time: 2hr 50min

Each design iteration:

- flow solver: 22s (MPI parallel)
- gradient calculation: 2min 47s (shared-memory parallel)
- geometry deformation: 4.4s (shared-memory parallel)

References I

- [1] Alexander Ward. “Aftbody design of winged-cone derived hypersonic vehicles”. PhD thesis. School of Mechanical and Mining Engineering, The University of Queensland, 2022.
- [2] Larry Schweikart. *The Hypersonic Revolution, Case Studies in the History of Hypersonic Technology: Volume III, the Quest for the Orbital Jet: The National Aero-Space Plane Program (1983-1995)*. Createspace Independent Publishing Platform, 1998. ISBN: 9781478146179.
- [3] B. Zhang et al. “Efficient Aerodynamic Shape Optimization of the Hypersonic Lifting Body Based on Free Form Deformation Technique”. In: *IEEE Access* 7 (Oct. 2019), pp. 147991–148003. DOI: 10.1109/ACCESS.2019.2945082.
- [4] T. W. Drayna. “Design and Optimization of Hypersonic Inward-Turning Inlets”. PhD thesis. The University of Minnesota, 2011.
- [5] Song Chen et al. “Aerodynamic Shape Optimization of Common Research Model Wing–Body–Tail Configuration”. In: *Journal of Aircraft* 53.1 (2016), pp. 276–293. DOI: 10.2514/1.C033328.
- [6] T. M. Leung and D. W. Zingg. “Aerodynamic Shape Optimization of Wings Using a Parallel Newton-Krylov Approach”. In: *AIAA Journal* 50.3 (2012), pp. 540–550. DOI: 10.2514/1.J051192.
- [7] G. K. W. Kenway. “A Scalable, Parallel Approach for Multi-Point, High-Fidelity Aerostructural Optimization of Aircraft Configurations”. PhD thesis. Graduate Department of Aerospace Engineering, University of Toronto, 2013.

References II

- [8] T. A. Reist et al. “Cross Validation of Aerodynamic Shape Optimization Methodologies for Aircraft Wing-Body Optimization”. In: *AIAA Journal* 58.6 (2020), pp. 2581–2595. DOI: 10.2514/1.J059091.
- [9] J. E. Hicken and D. W. Zingg. “Aerodynamic Optimization Algorithm with Integrated Geometry Parameterization and Mesh Movement”. In: *AIAA Journal* 48.2 (2010), pp. 400–413. DOI: 10.2514/1.44033.
- [10] P. E. Gill, W. Murray, and M. A. Saunders. “SNOPT: An SQP Algorithm for Large-Scale Constrained Optimization”. In: *SIAM Review* 47.1 (2005), pp. 99–131. DOI: 10.1137/S0036144504446096.