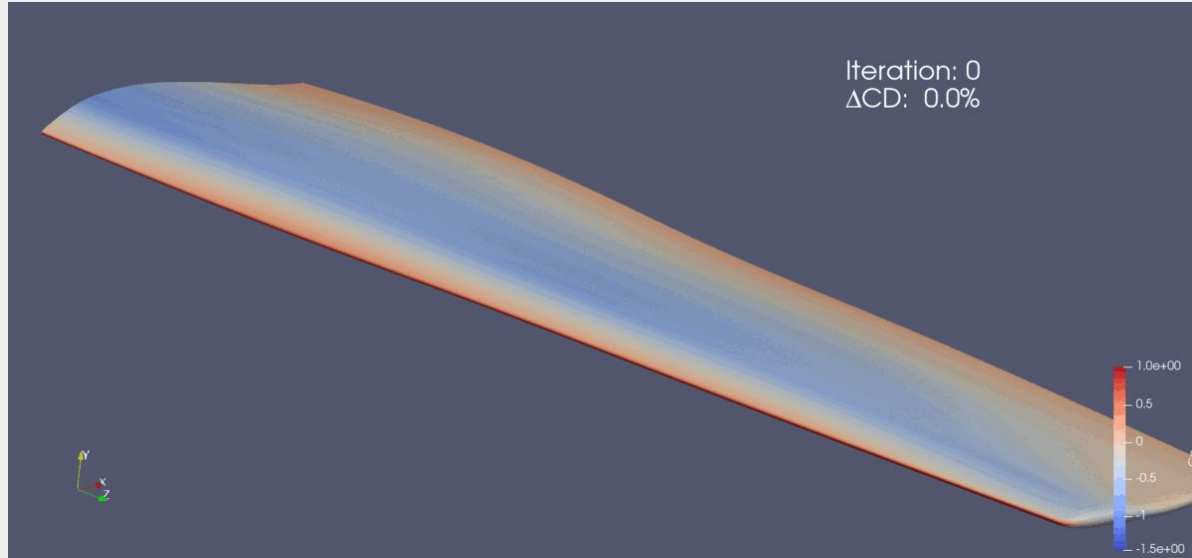


# Quantifying the effect of adding planform variables to a UAV wing aerodynamic optimization



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7/28/22

# Introduction

- The aerospace engineering industry typically uses manual design considerations to optimize wing performance.
- Computational optimization methods can potentially be more accurate than using conventional experience.
- Eventual dependence on computational optimization will make the design testing process more physically accessible and less costly.
- UAVs are futuristic technologies used for aerial imagery, inspections, and mapping.



*NASA Dresden's F-15A  
Remotely Piloted Research  
Vehicle (RPRV)*

# Wing Geometry

$$AR = \frac{b^2}{S}$$

Aspect Ratio Formula

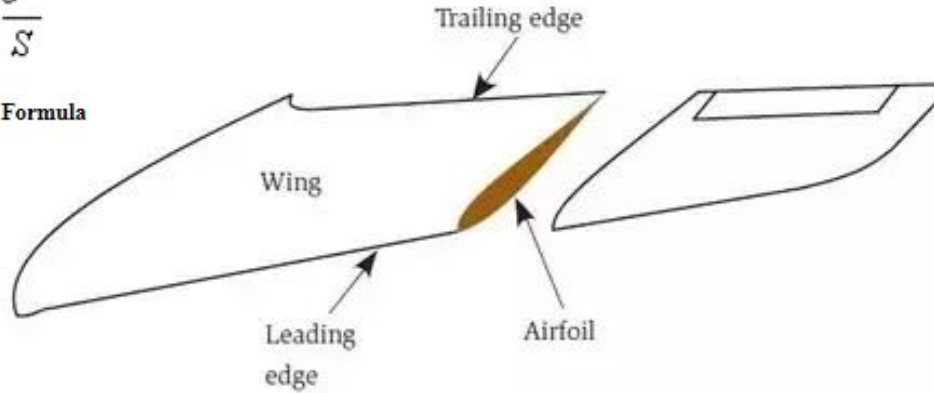


Fig. 1.3. The wing and airfoil.

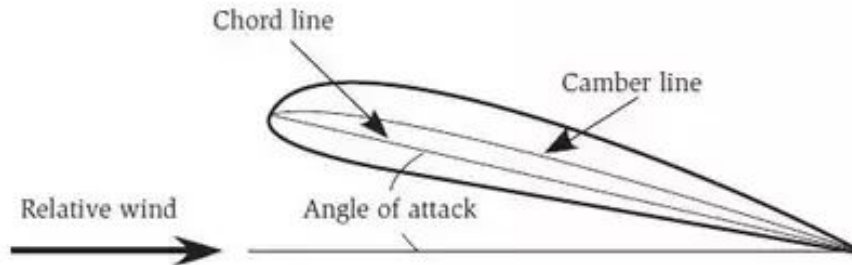
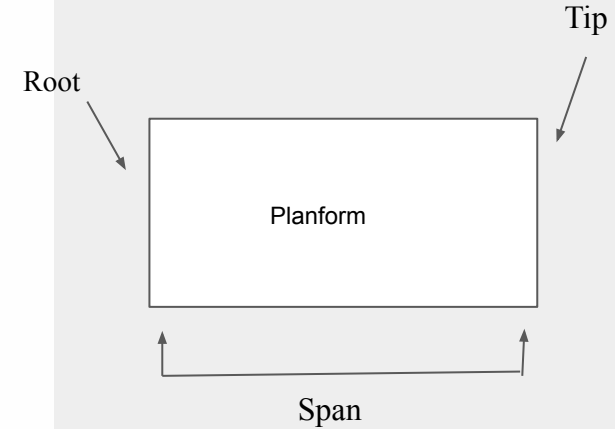


Fig. 1.4. Airfoil nomenclature.

$$\text{Taper Ratio } \lambda = \frac{\text{tip chord}}{\text{root chord}} = \frac{C_t}{C_r}$$



# Research Objective

Research question: Will the addition of planform variables to a shape-only case increase or decrease the drag of an optimized airfoil?

This study allows for the determination of variables that result in the largest drag reduction. The potential impact on the aerospace industry is great as it allows for concentrated attention on key variables, reducing wasted resources.

Objective function: Drag coefficient

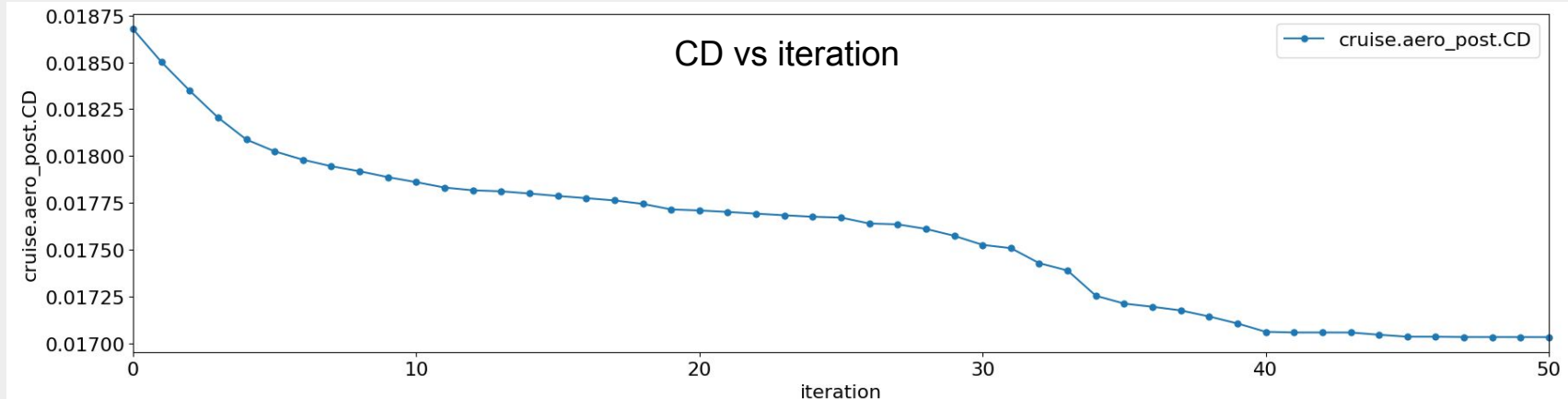


Figure 2: CD vs Iteration Plot

# Software Framework

The DA Foam (Discrete Adjoint with OpenFOAM for High-fidelity Multidisciplinary Design Optimization) software can automatically optimize an existing design according to given parameters and output a percentage of the design's performance increase.

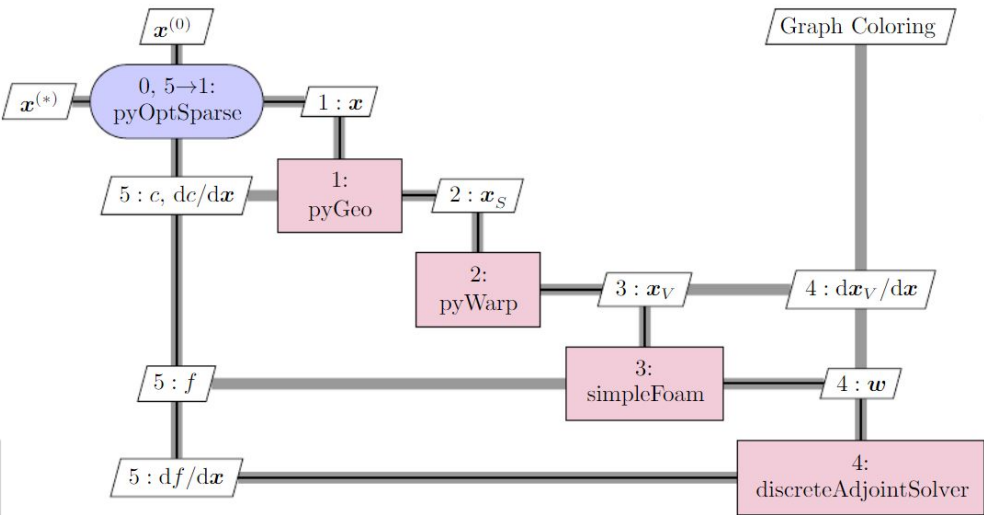


Figure 3: DA Foam Design Matrix

pyOptSparse	variable input (SNOPT)
pyGeo	surface geometry (FFD)
pyWarp	mesh deformation
simpleFoam	CFD solver (OPENFoam)
discreteAdjointSolver	DA Foam

Table 1: DA Foam Modules

# Computation Process

- An HPC (high performance computer) was used to lessen computational time.
- Used Stampede2, a Linux-based UTexas Austin supercomputer.
- ICX-normal computing nodes for shorter queue times.
- Cases were each run on 4 nodes (computers), 80 cores/node. 320 tasks (commands)/case.



Figure 4:  
Stampede2

```
#!/bin/bash
#SBATCH -A TG-ATM140019      # project ID
#SBATCH -J planform_12      # job name
#SBATCH -o log-%j.txt       # output and error file name (%j expands to jobID)
#SBATCH -n 320              # total number of mpi tasks requested
#SBATCH -N 4                # total number of nodes
#SBATCH -p icx-normal       # queue (partition) -- normal, development, etc.
#SBATCH -t 10:00:00         # run time (hh:mm:ss)
#SBATCH --mail-type=ALL      # setup email alert
#SBATCH --mail-user=ppak@bu.edu
```

Figure 5: HPC submission .sh script

# Python Interface

- runscript.py is a python script defining parameters, variables, tolerances, and constraints for a given case.
- Tolerances and parameters were kept constant for all cases.

Parameter	Variable	Value	Unit
Freestream Velocity	U0	100.0	m/s
Freestream Pressure	p0	101325.0	kPa
Temperature	T0	300.0	K
Target Lift Coefficient	CL_target	0.375	
Angle of attack	aoa0	2.0	deg
Planform area	A0	3.0	m^2
Mach Number	M	0.3	
Reynolds Number	Re	2.20E+05	
Span	span0	3.0	m
Chord	chord0	1.0	m

Table 2: runscript.py parameters

# Case Breakdown

Case	VarTypes	Constraints	Total Variables	Total Constraints
Shape	aoa, twist, shape	thickcon, volcon, lecon, telcon, CL	126	764
Planform 1	aoa, taper, span	volcon, CL	4	3
Planform 2	aoa, taper	CL	3	2
Planform 3	aoa, span	CL	2	2
Combination	Aoa, twist, shape, taper, span	thickcon, volcon, lecon, telcon, CL	129	764

Table 3: Case Descriptions

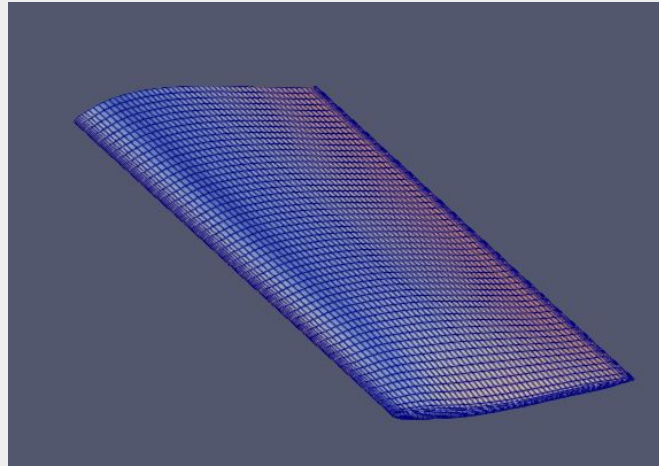


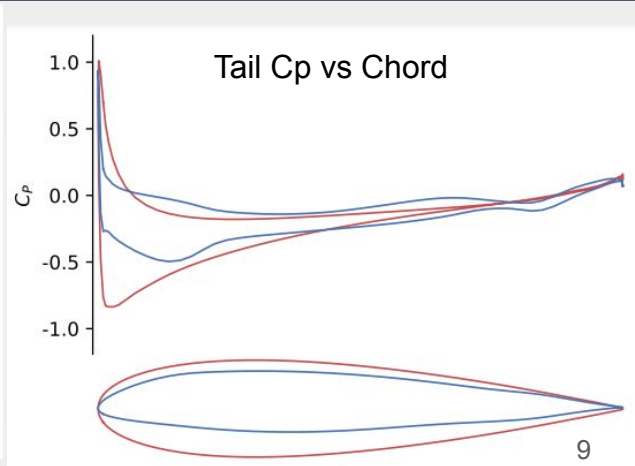
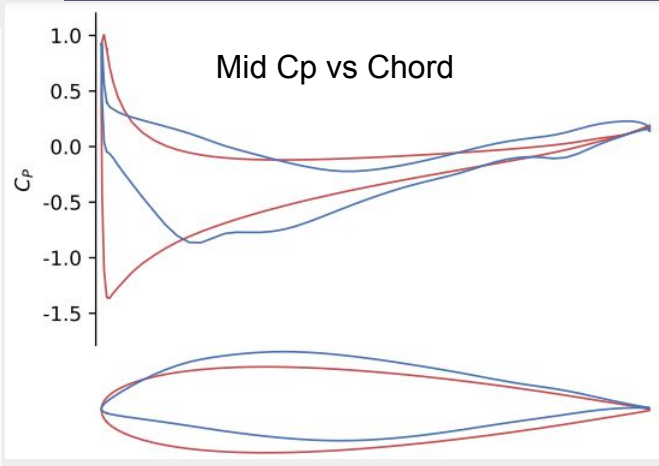
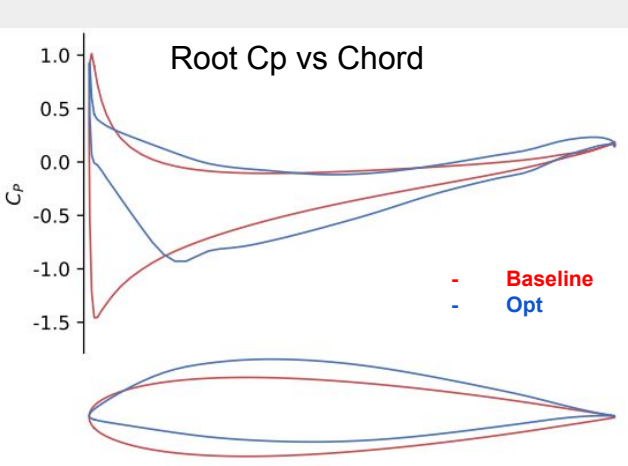
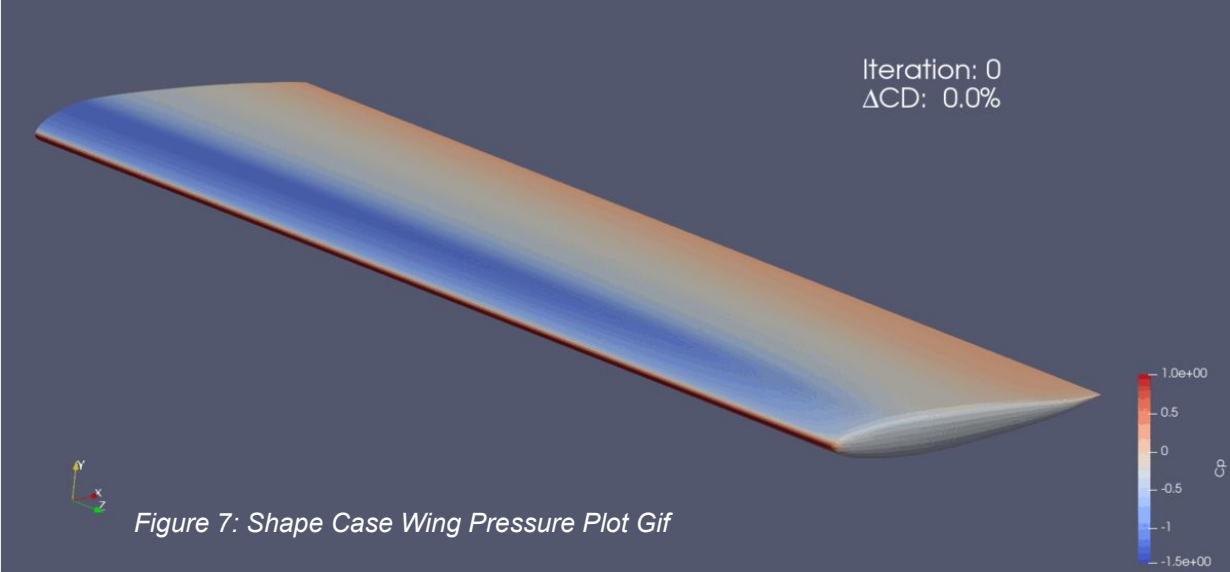
Figure 6: Shape Case Wing Mesh



# Shape Case Results

Drag reduction	8.79%
Runtime	16:08:46
Aspect Ratio	3
Taper ratio	1

Table 4: Shape Case Results



# Planform Case 1 Results

Drag reduction	11.50%
Runtime	2:41:44
Aspect Ratio	5.57
Taper ratio	1

Table 5: Planform Case 1 Results

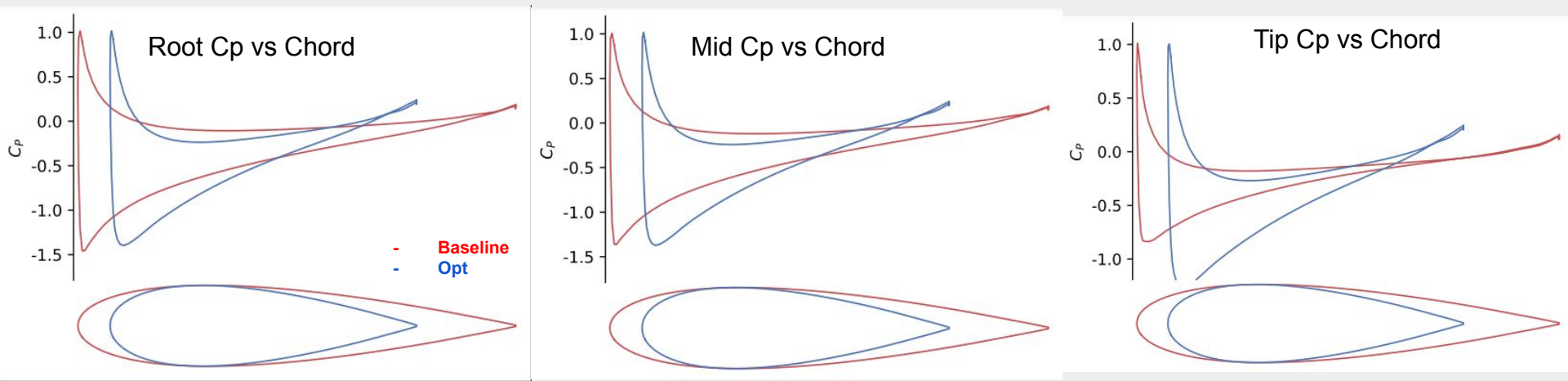
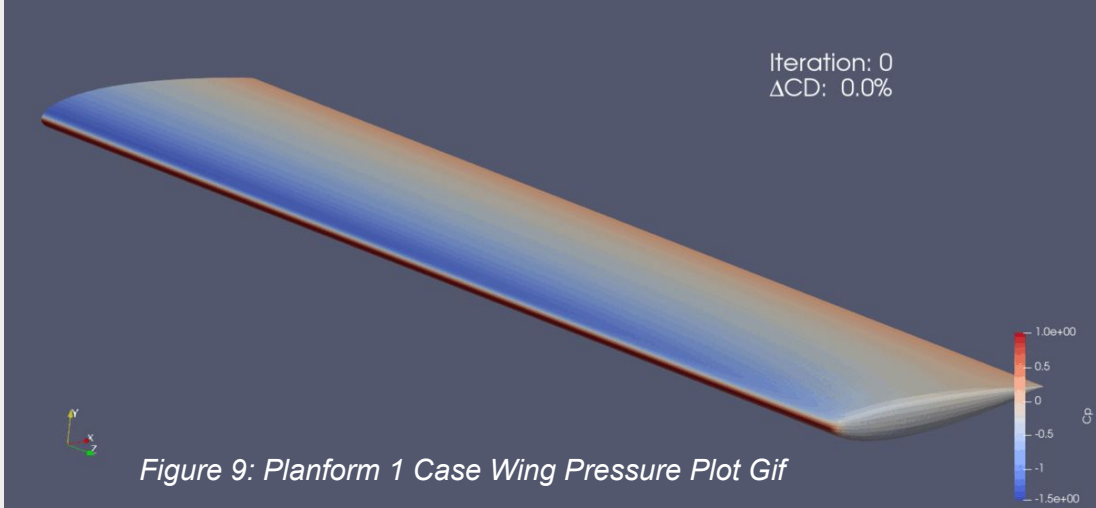


Figure 10: Planform Case 1 Plots

# Planform Case 2 Results

Drag reduction	8.47%
Runtime	4:43:05
Aspect Ratio	2
Taper ratio	0.5

Table 6: Planform Case 2 Results

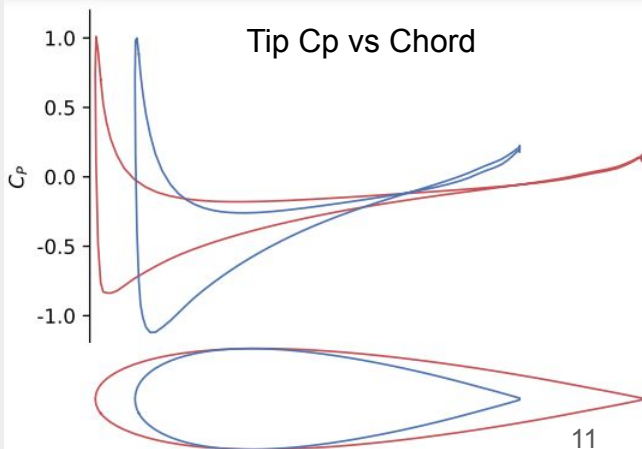
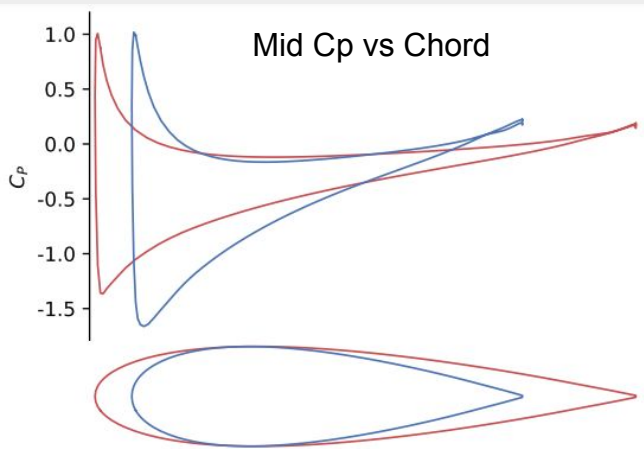
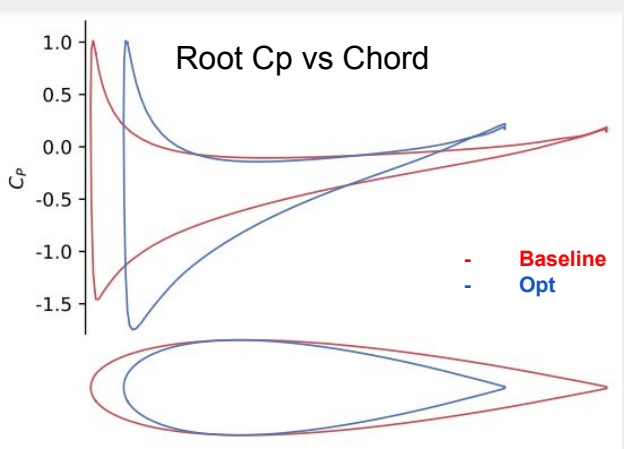
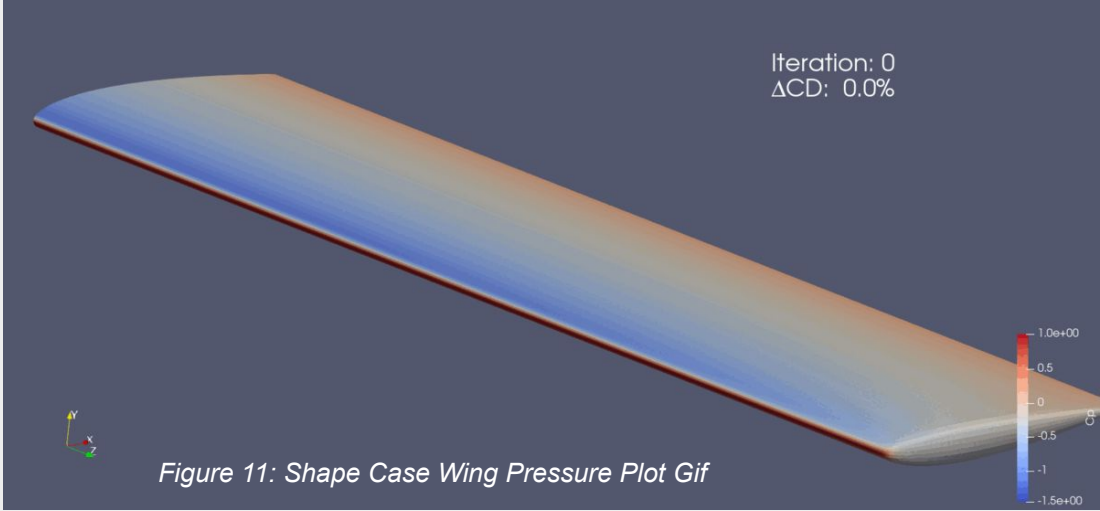


Figure 12: Planform Case 1 Plots

# Planform Case 3 Results

Drag reduction	3.49%
Runtime	3:35:14
Aspect Ratio	3.59
Taper ratio	1

Table 7: Planform Case 3 Results

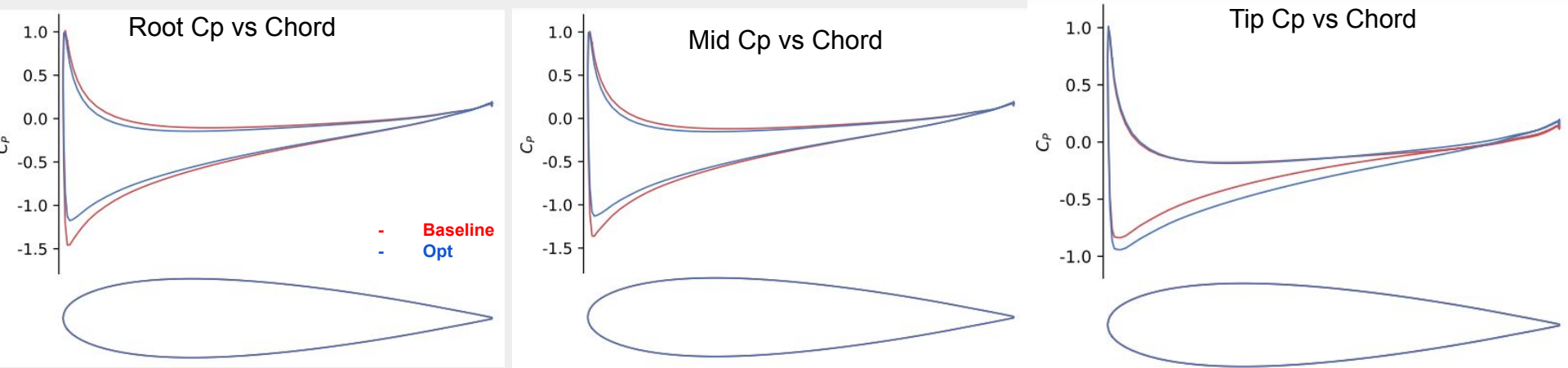
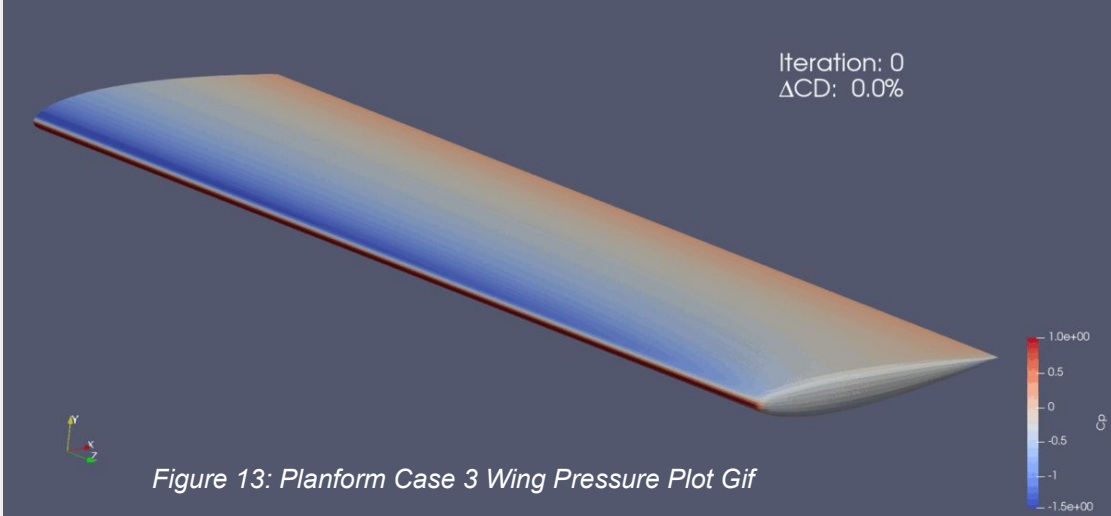


Figure 14: Planform Case 3 Graphs

# Combined Case Results

Drag reduction	17.02%
Runtime	12:16:12
Aspect Ratio	3.68
Taper ratio	0.89

Table 8: Combined Case Results

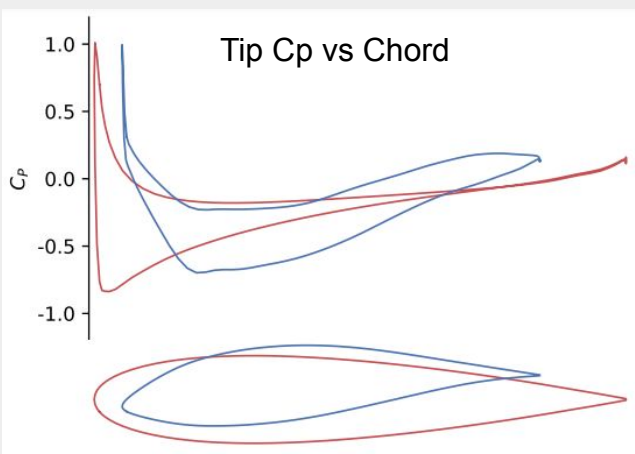
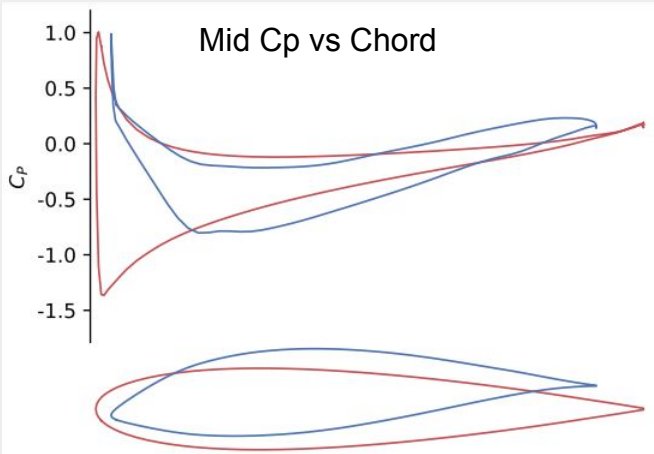
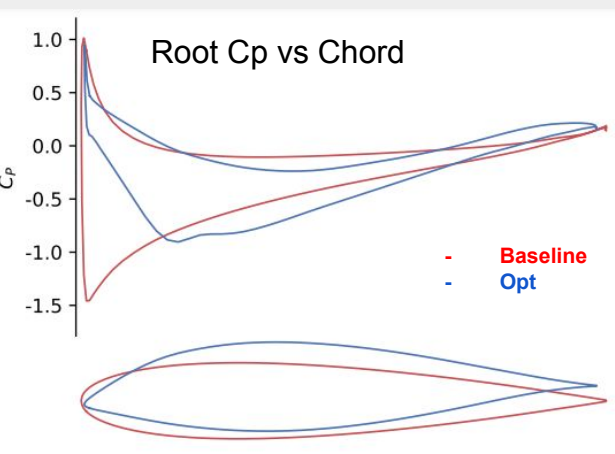
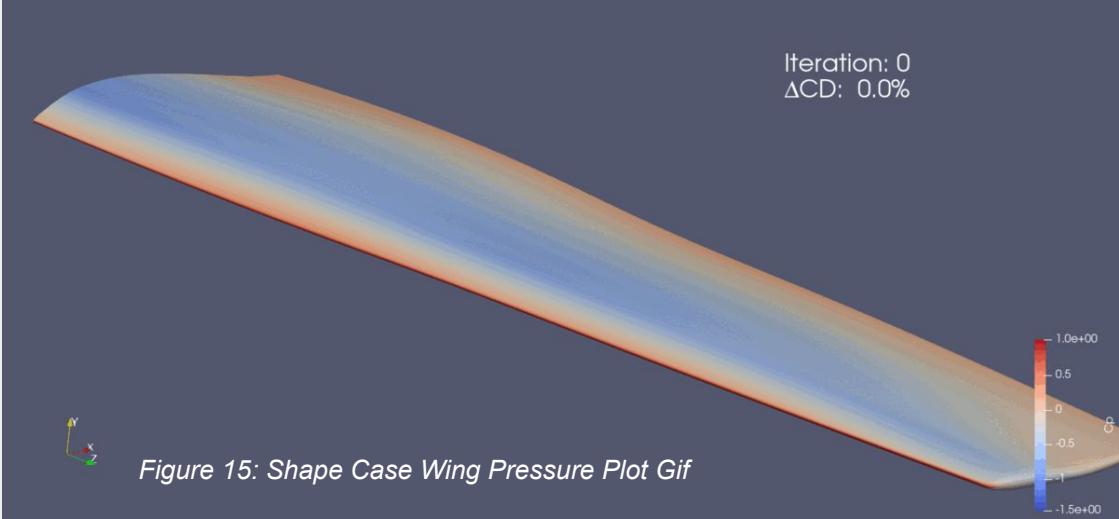


Figure 16: Combined Case Graphs

# Conclusions

- The addition of planform variables to a shape case will further decrease the drag by 8.23%.
- As drag decreases, taper ratio decreases and aspect ratio increases.
- The most time-efficient case was the planform 1 case.
  
- **Verification:** Wind tunnel testing.
- The consistency of the results can be determined using other airfoil shapes and running cases with various flight conditions.
  
- **Future studies:** Account for stalling using ailerons.
- Consider weight to account for induced drag from a longer span.

# Acknowledgements

This research was funded by the NSF LAUNCH-UAS program.

My research could have not been accomplished without Dr. Ping He, who provided extensive mentorship and support throughout the program.

I am sincerely grateful to Heyecan Utke, who dedicated extra effort towards the post-processing of my simulations and was consistently available for guidance.

I warmly thank Yiqi Liang, who enabled the successful completion of my research through helpful comments and support.