



# S.M.A.C.

(Smart Models for Automated Control)

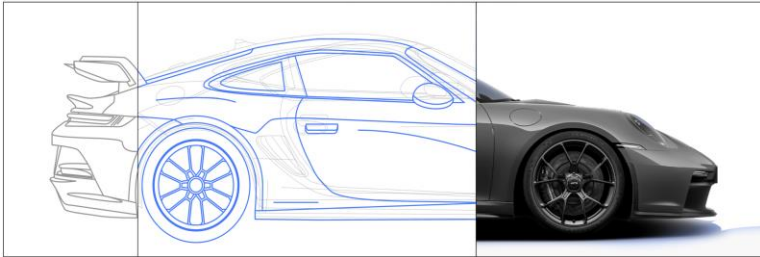
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Zhen Wei  
Project PI: Pascal Fua

Swiss JRC  
Workshop

# Computer Assisted Engineering

Machines need to be designed for:

- Performance
- Durability
- Ease of control
- .....



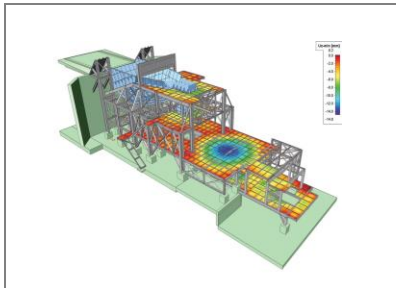
Automotive



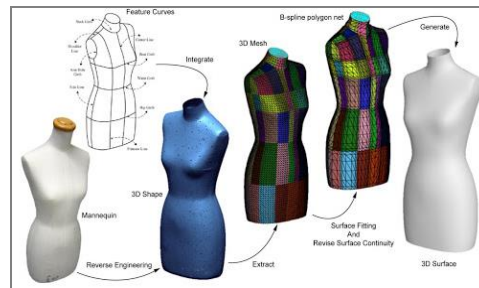
Space



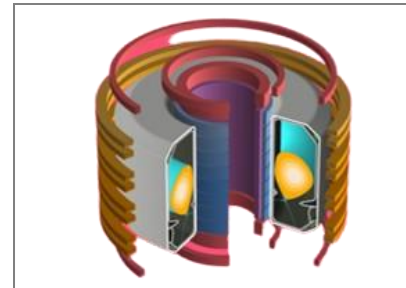
Electrification



Architecture

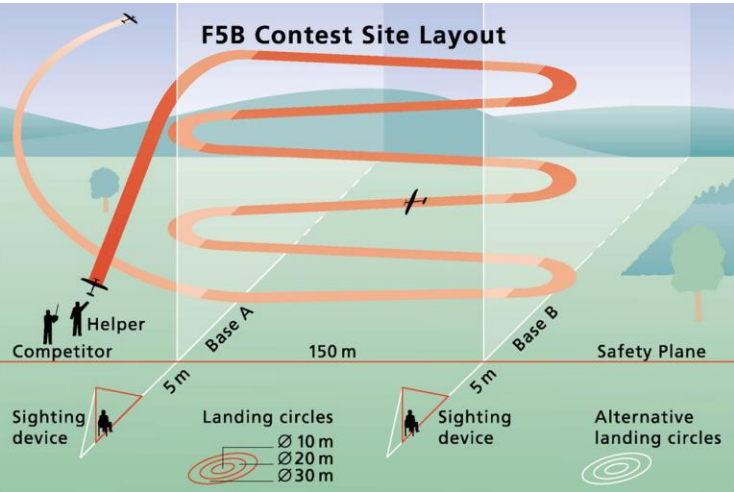


Textile



Nuclear fusion

# EPFL Long-Term Objective



- Designing a glider to:
- perform a complex task
  - be easy to control
  - be structurally sound

- F5B Electric Glider Competition

# EPFL Advantages

## ▪ Conventional:

- 1 Operating Point
- 1 Performance Metric
- Computationally Expensive
- Labour Intensive
- Not Reusable

>6-7h



## ▪ Our Approach:

- Whole Trajectory
- Adaptable Performance Metric
- Training is Expensive
- Inference is ⚡
- Reusable

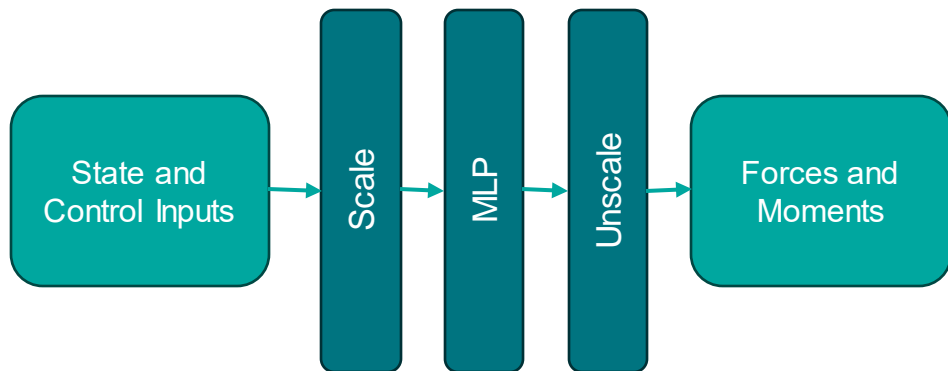
<1s





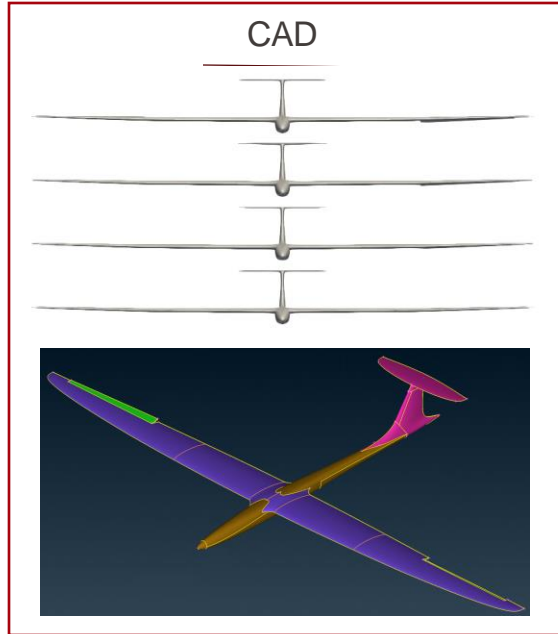
Given an existing glider:

1. Build a dataset and train a Deep Network to predict its dynamics
2. Compute optimal trajectories for its current shape
3. Optimize the shape to further improve the trajectories



- Two supervision sources to train this model:
  - **Simulation data** to predict the response of the aircraft to given control inputs
  - **Wind tunnel data** for greater accuracy

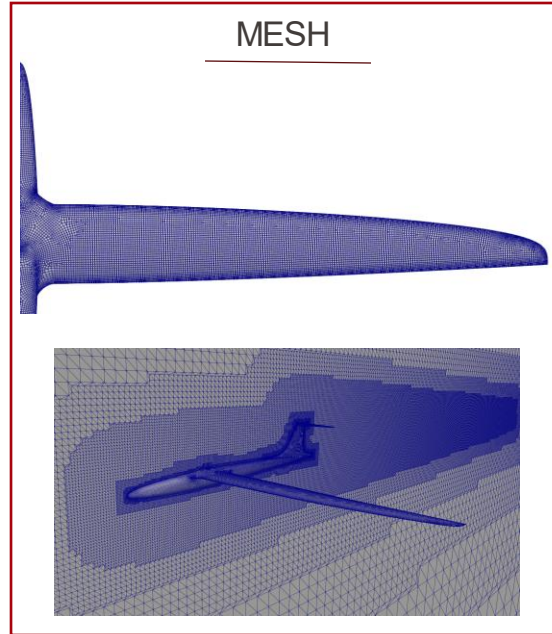
⚠ There might be inconsistencies between the **two**



**Rhino / Ansa**

~1-3h

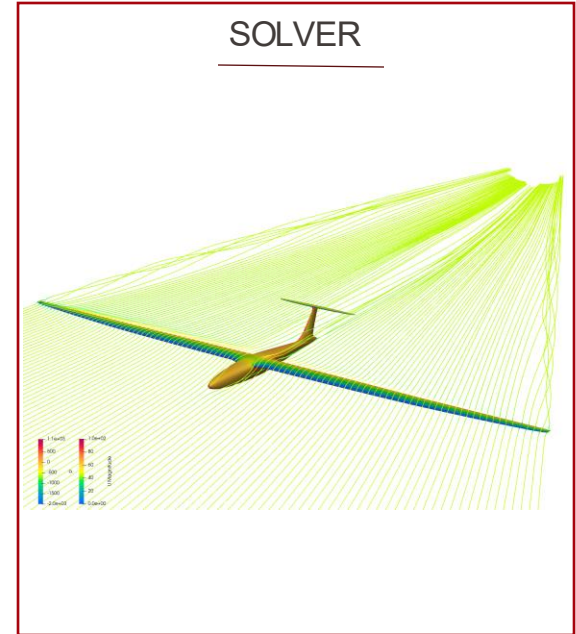
- Different configurations
- Surface identifications and cleaning
- General preprocessing for meshing



**Ansa**

~1-4h

- Automated batch mesh workflow
- Anisotropic mesh on L.E. & T.E.
- Wake refinement

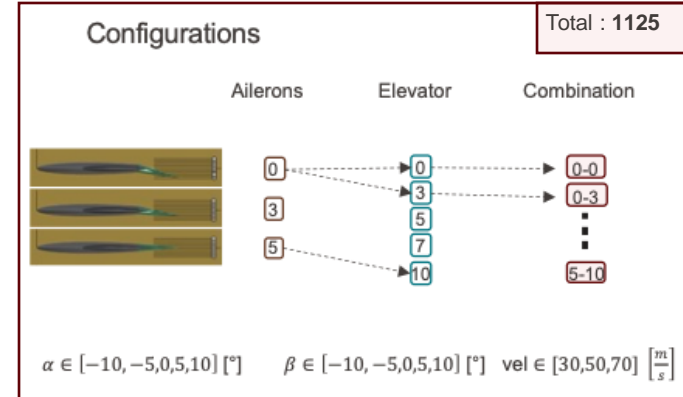
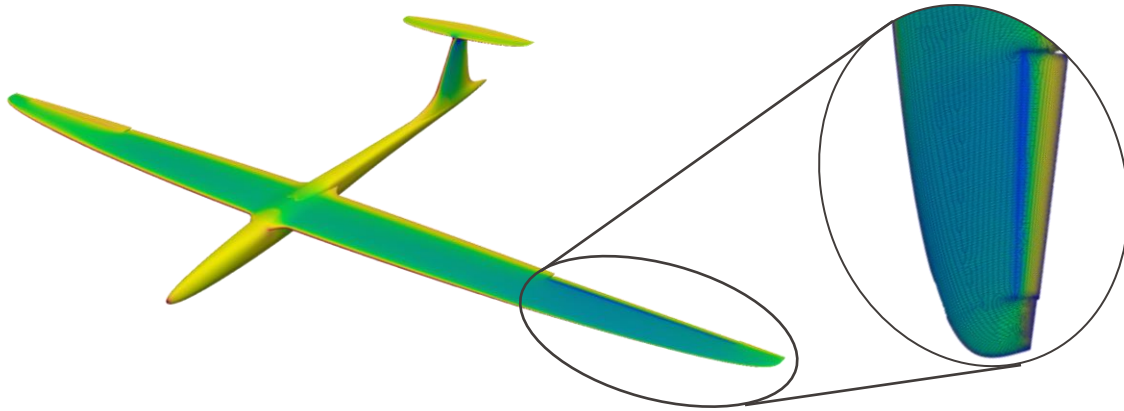


**OpenFOAM**

>5h

- RANS, Spallart-Allmaras
- Automated B.C. changes
- Output forces, moments, pressure

# EPFL Deflecting the Controls



Number of CPU core per simulation: 32 / Average run time: ~5h30



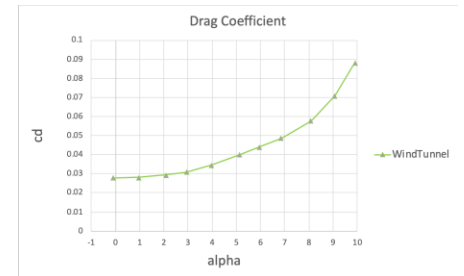
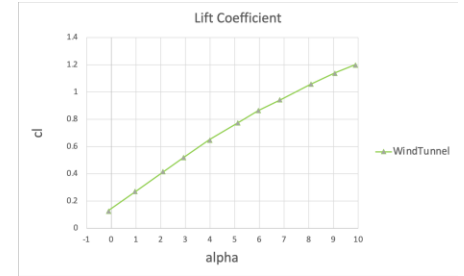
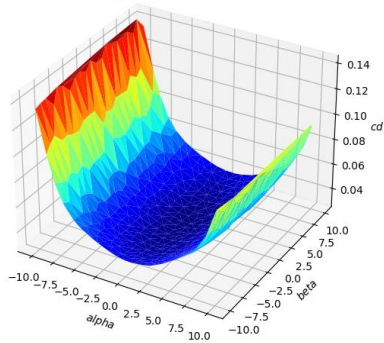


S. Seim



S. Seim

drag coefficient @ 49 mps



University of Washington Aeronautical Laboratory

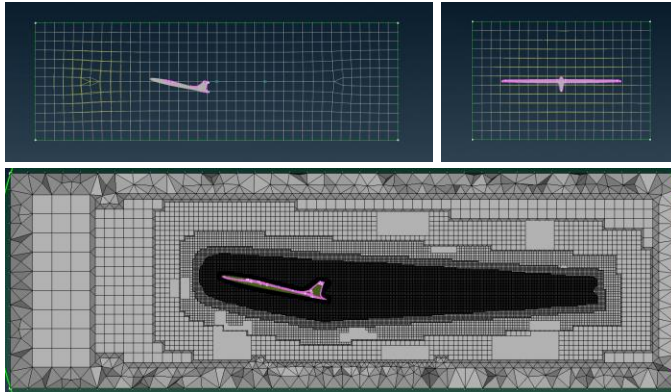
Windtunnel



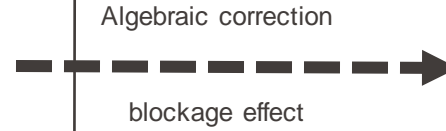
Mesh check



Data merging



Freestream

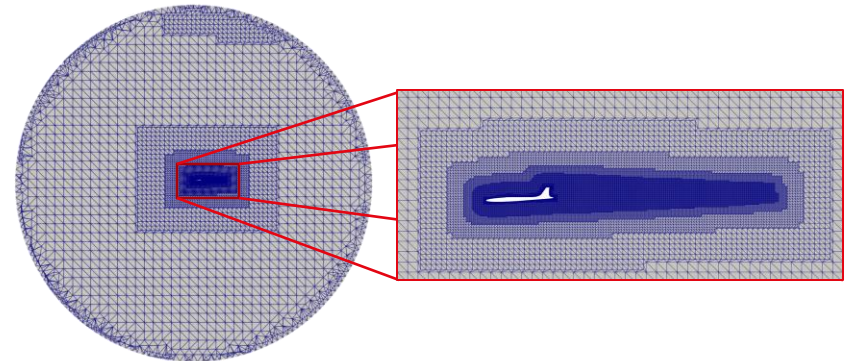


Experimental data  
Freestream

Mesh check

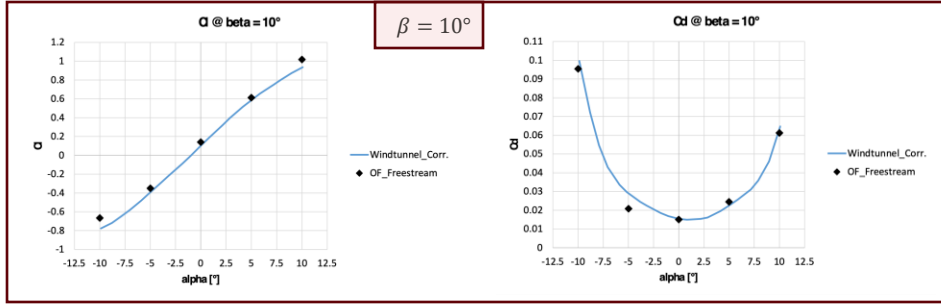
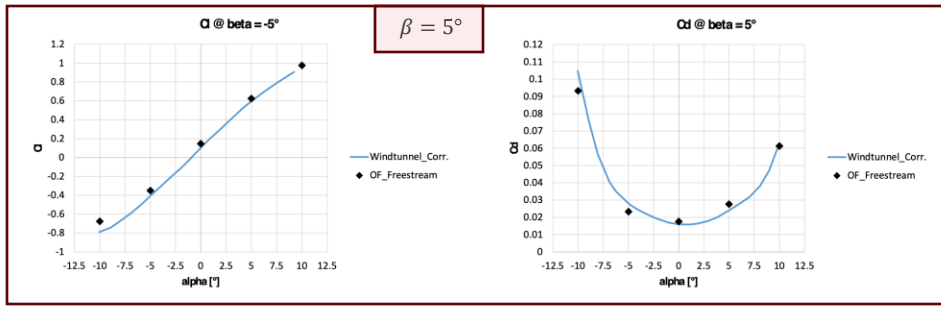
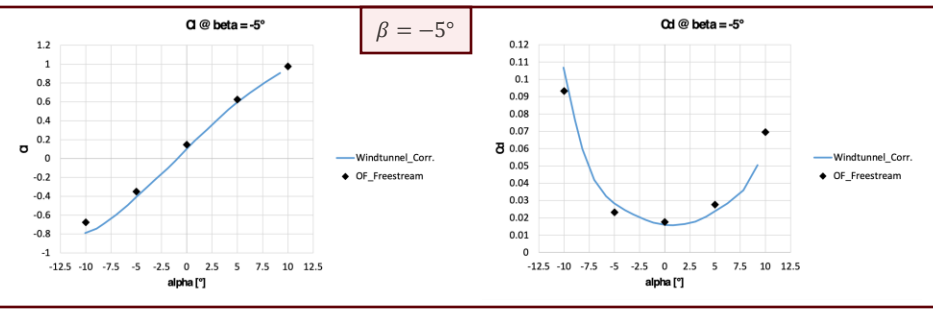
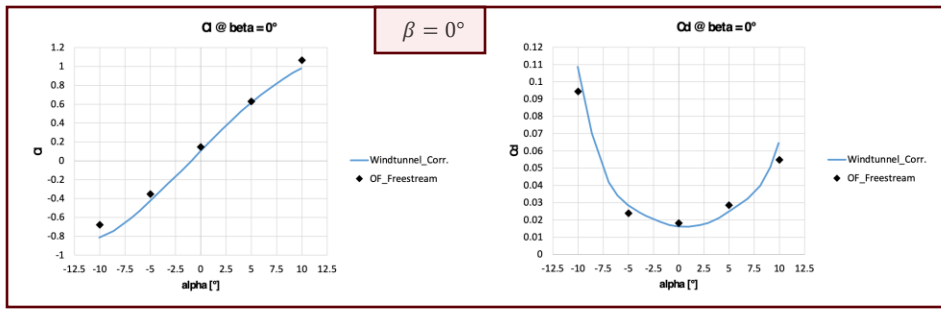
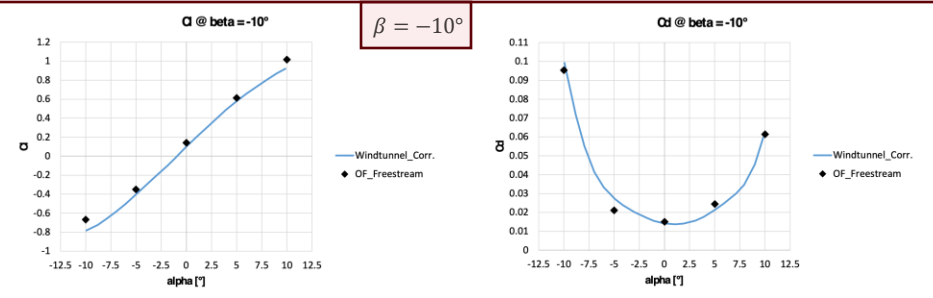


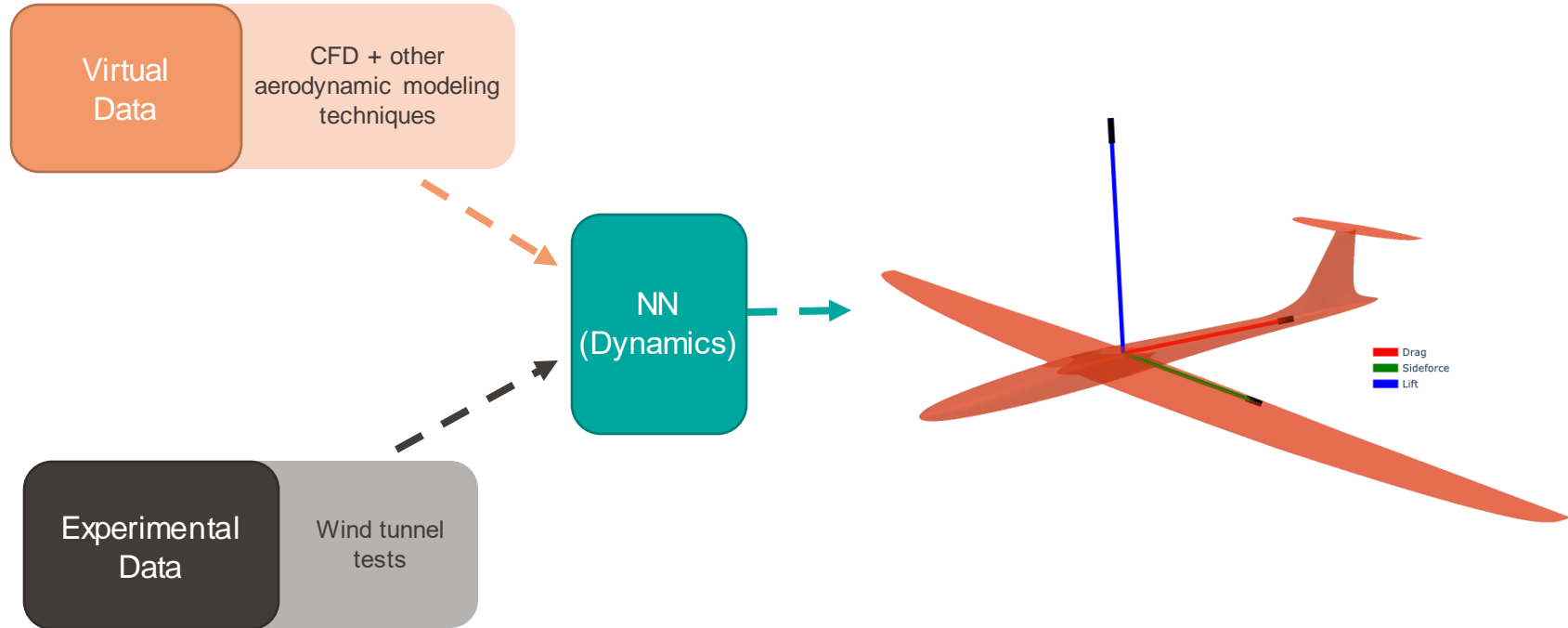
Data merging

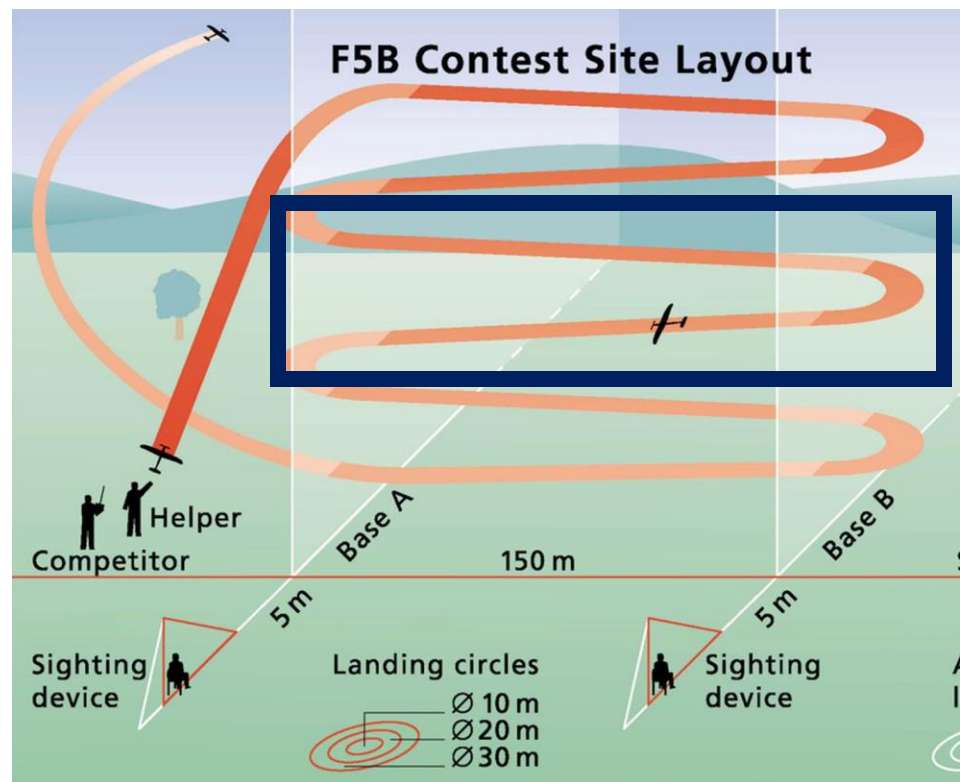
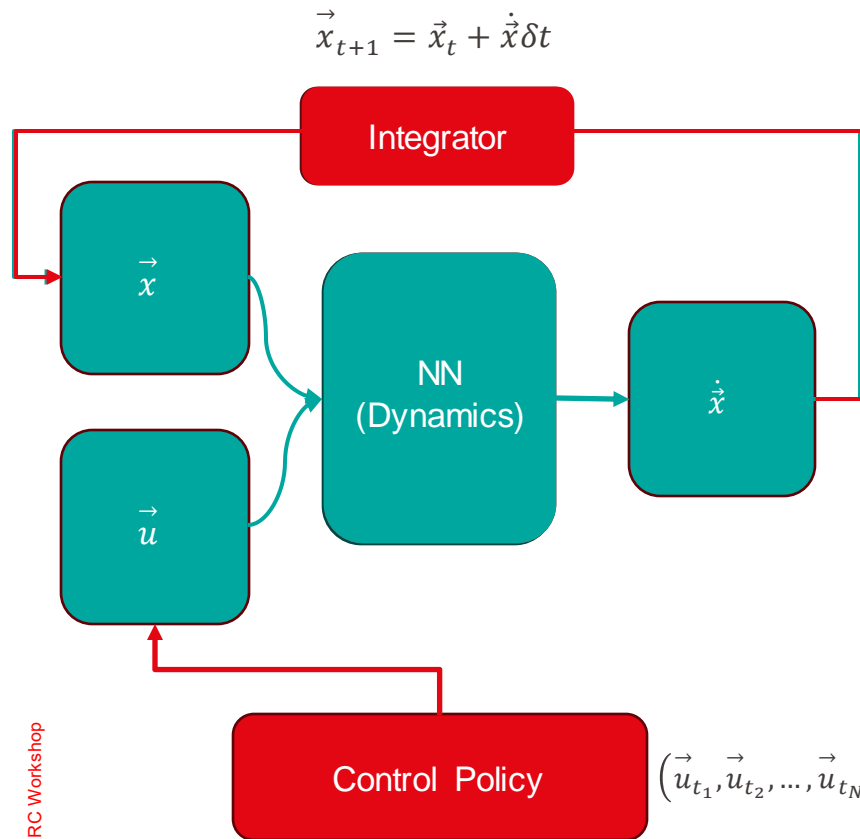


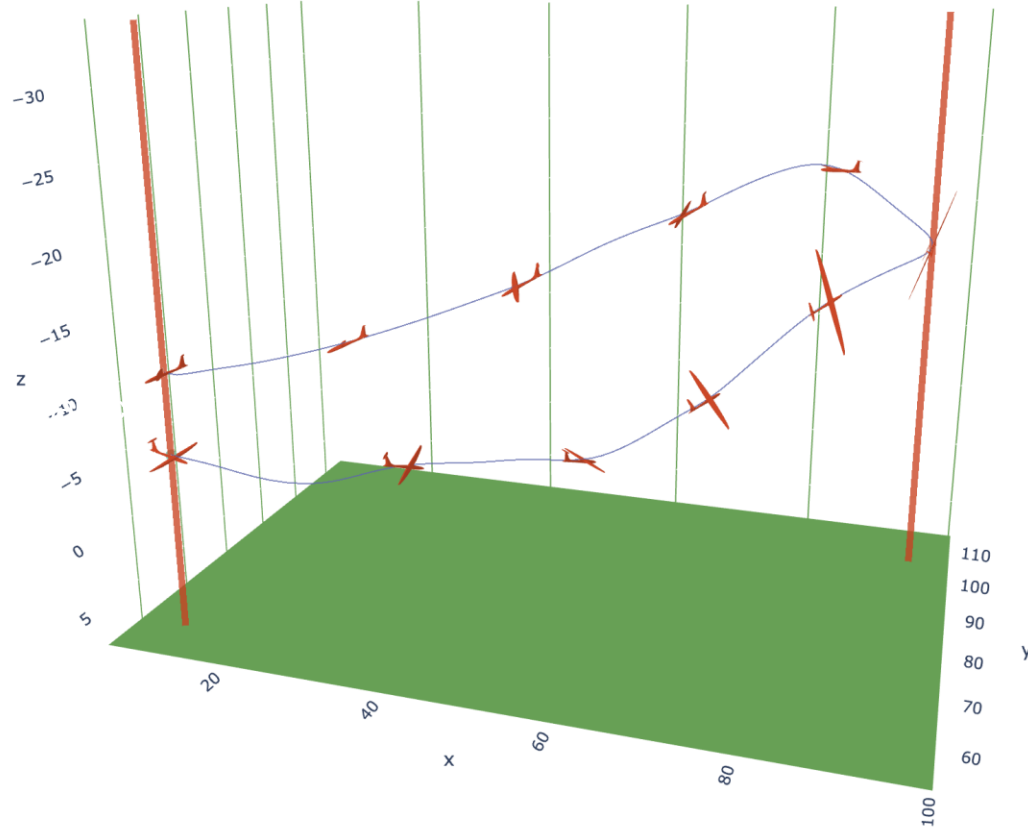
# EPFL Freestream Results

## Experimental / OpenFOAM

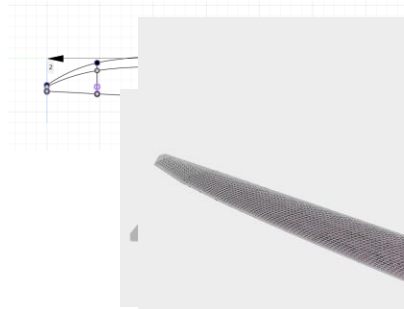
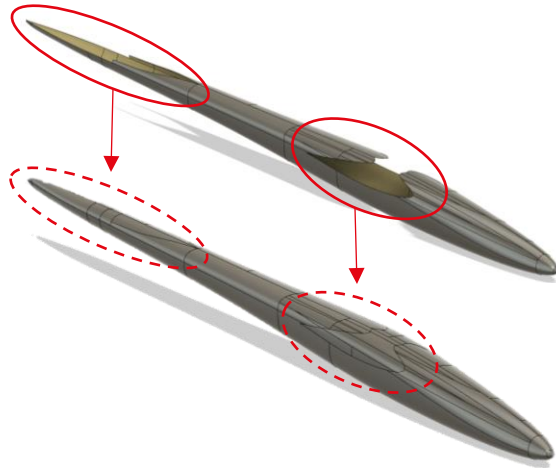
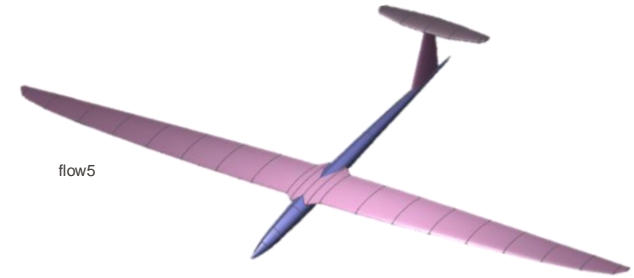
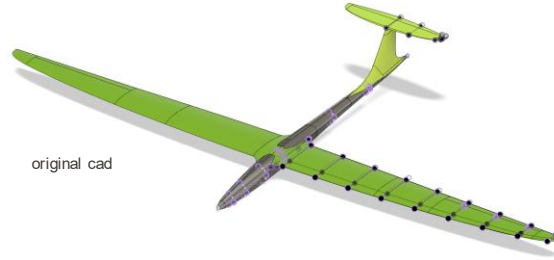




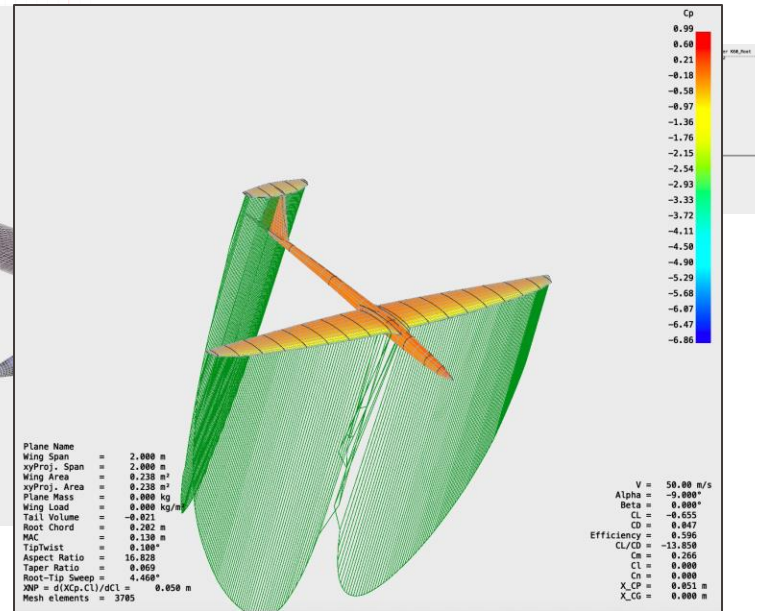




## Glider reverse engineering



Plane Name	=	2.000 m
Wing Span	=	2.000 m
xyProj. Area	=	0.238 m²
Wing Area	=	0.238 m²
Plane Mass	=	0.000 kg
Wing Load	=	0.000 kg/m²
Tail Volume	=	-0.021
Root Chord	=	0.262 m
MAC	=	0.130 m
TipTwist	=	0.100°
Aspect Ratio	=	10.020
Taper Ratio	=	0.069
Root-Tip Sweep	=	4.468°

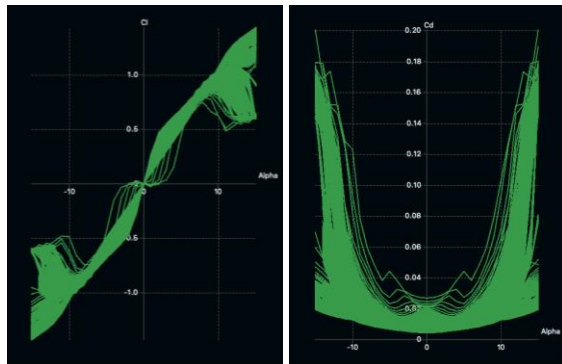
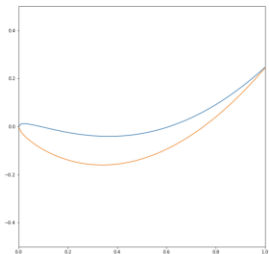


Run time for ~200 points = 5min !

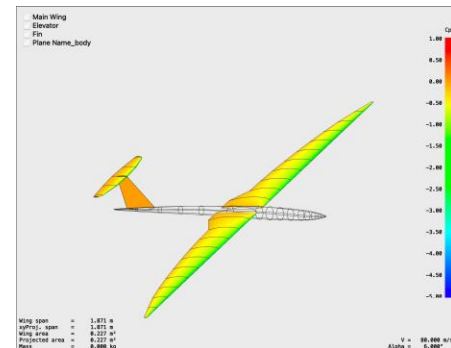
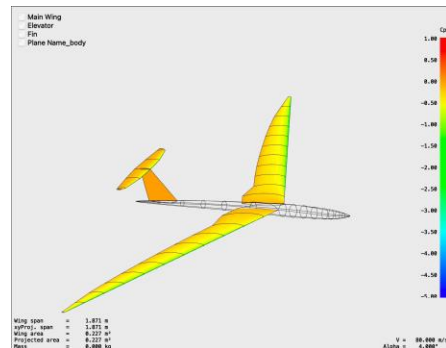
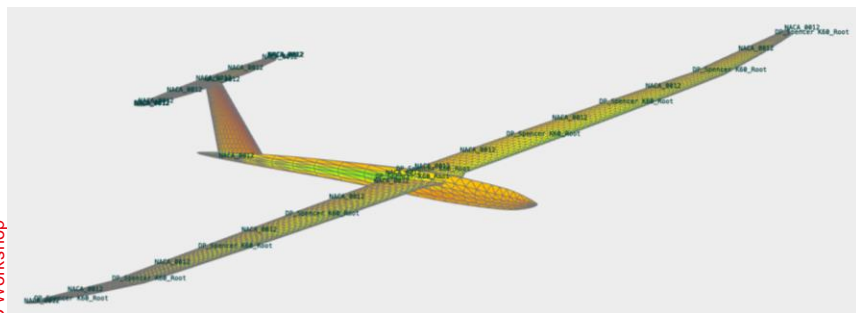
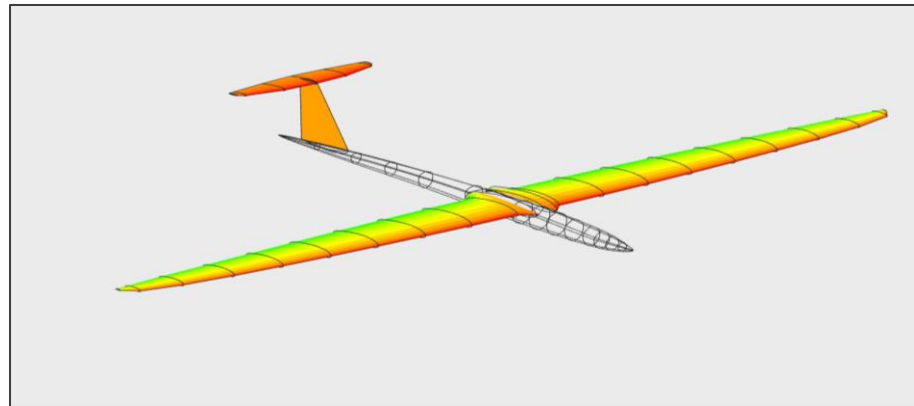


## Flow5

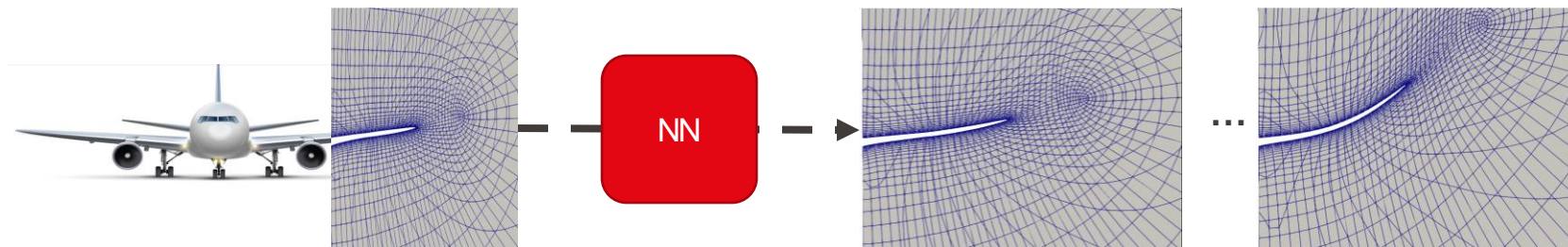
### NACA airfoil (Python)



## 3D deformation







We are building a deep geometric learning model to:

- manipulate the aircraft's geometry
- deform the computational meshes accordingly for simulation purposes

