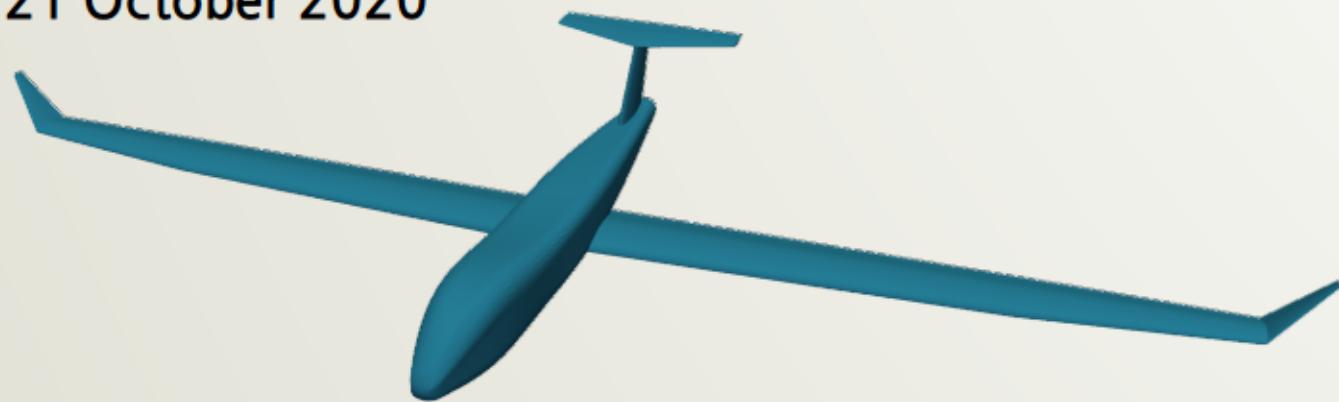


READ/EWADE 2020



Using surrogate models to speed up the creation of aerodynamic databases in CEASIOMpy

21 October 2020



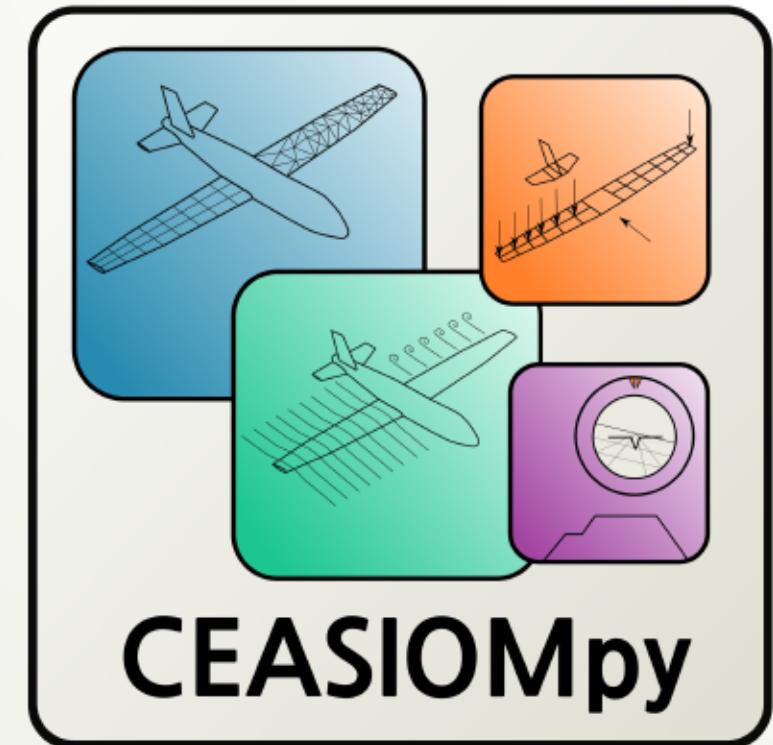
Aidan Jungo
Vivien Riolo
Jan Vos

CFS Engineering
Computational Fluids & Structures Engineering

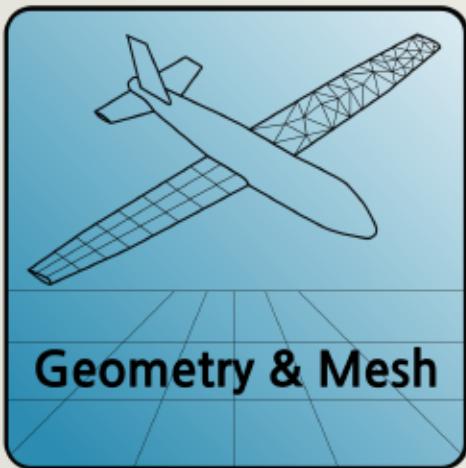
AGILE^{4.0}

Introduction

- Conceptual aircraft design environment
- Legacy from FP6 SimSAC (2006-2009)
- Based on Python 3.5 and CPACS 3.1
- Open-source:
 - <http://github.com/cfsengineering/CEASIMpy>
 - <http://ceasiompy.readthedocs.io>



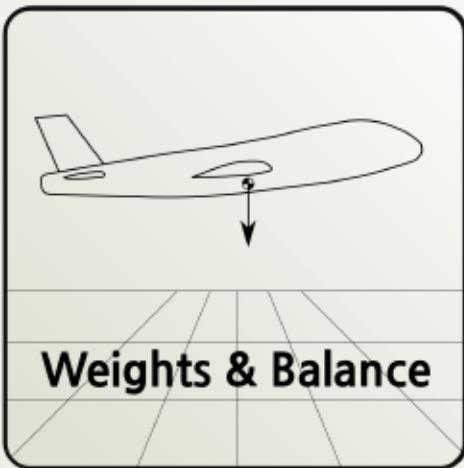
Introduction



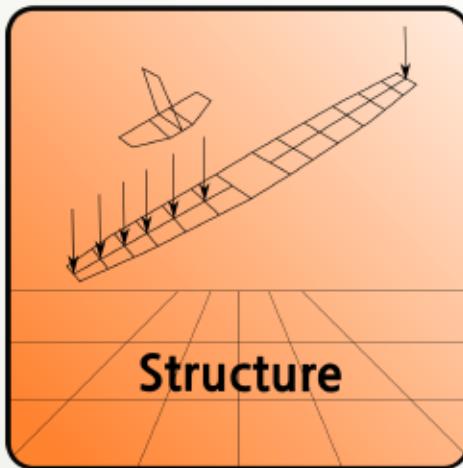
Geometry & Mesh



Aerodynamics



Weights & Balance



Structure

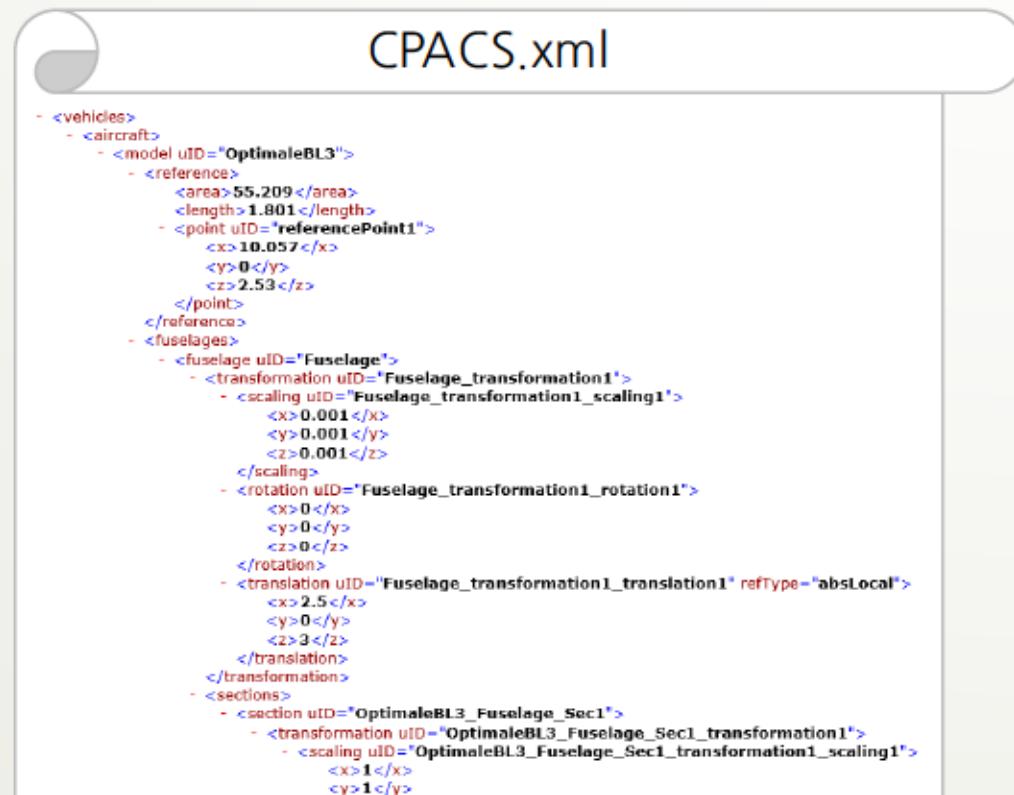


Mission Analysis

- CPACSCreator
- CLCalculator
- SUMO
- PyTornado
- SU2
- SkinFriction
- Weights conventional
- Weights unconventional
- Balance conventional
- Balance unconventional
- AeroFrame
- FrameAT
- Range
- Static Stability

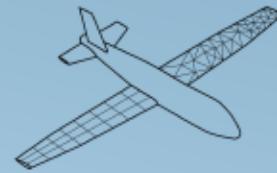
Introduction

- CPACS format
 - Common Parametric Aircraft Configuration Schema
 - Developed by the DLR (Germany)
 - XML based
 - Libraries TiXI/TiGL
- Can be stored:
 - Geometry
 - Structure
 - Weights
 - Aerodynamics databases
 - ...



The screenshot shows a code editor window with the title "CPACS.xml". The XML code is displayed in a hierarchical tree view, where each node is represented by a small circular icon and its corresponding XML tag. The code itself is as follows:

```
<vehicles>
  <aircraft>
    <model uID="OptimaleBL3">
      <reference>
        <area>55.209</area>
        <length>1.801</length>
        <point uID="referencePoint1">
          <x>10.057</x>
          <y>0</y>
          <z>2.53</z>
        </point>
      </reference>
      <fuselages>
        <fuselage uID="Fuselage">
          <transformation uID="Fuselage_transformation1">
            <scaling uID="Fuselage_transformation1_scaling1">
              <x>0.001</x>
              <y>0.001</y>
              <z>0.001</z>
            </scaling>
            <rotation uID="Fuselage_transformation1_rotation1">
              <x>0</x>
              <y>0</y>
              <z>0</z>
            </rotation>
            <translation uID="Fuselage_transformation1_translation1" refType="absLocal">
              <x>2.5</x>
              <y>0</y>
              <z>3</z>
            </translation>
          </transformation>
          <sections>
            <section uID="OptimaleBL3_Fuselage_Sec1">
              <transformation uID="OptimaleBL3_Fuselage_Sec1_transformation1">
                <scaling uID="OptimaleBL3_Fuselage_Sec1_transformation1_scaling1">
                  <x>1</x>
                  <y>1</y>
                </scaling>
```

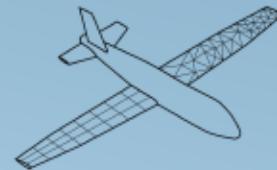


Geometry & Mesh

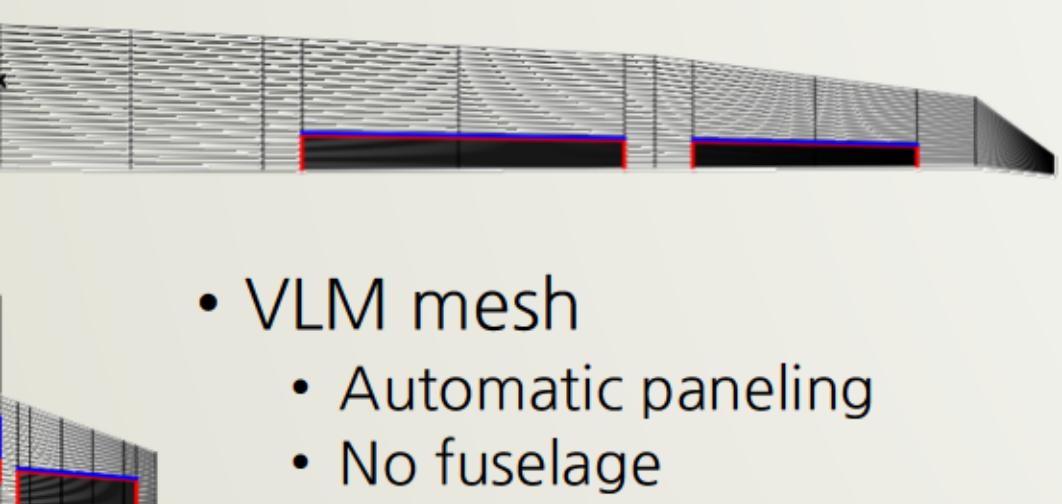
- OPTIMALE aircraft
 - UAV
 - Medium Altitude Long Endurance
 - German AeroStruct research project
 - H2020 - AGILE4.0

| Semi-span | 16,2 m |
|-----------------|-------------------|
| Ref. area | 55 m ² |
| Cruise speed | 150 m/s |
| Cruise altitude | 15'500 m |

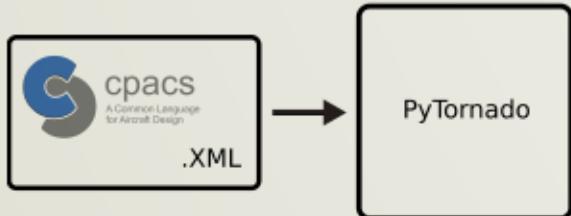




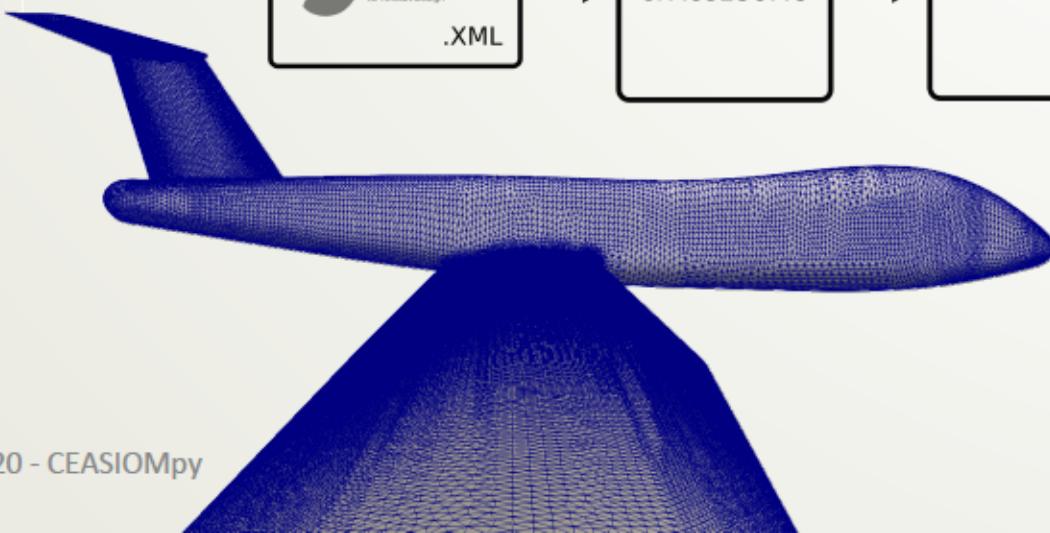
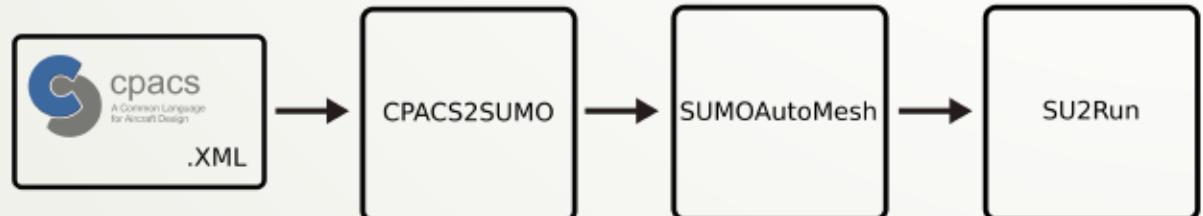
Geometry & Mesh



- VLM mesh
 - Automatic paneling
 - No fuselage

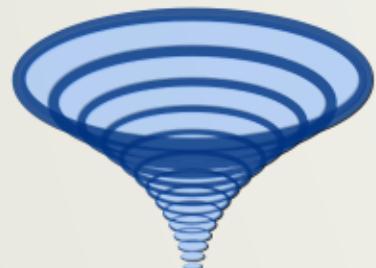


- Unstructured tetrahedron mesh
 - SUMO/Tetgen
 - Euler equation
 - Mesh convergence has been tested





Aerodynamic



PyTornado

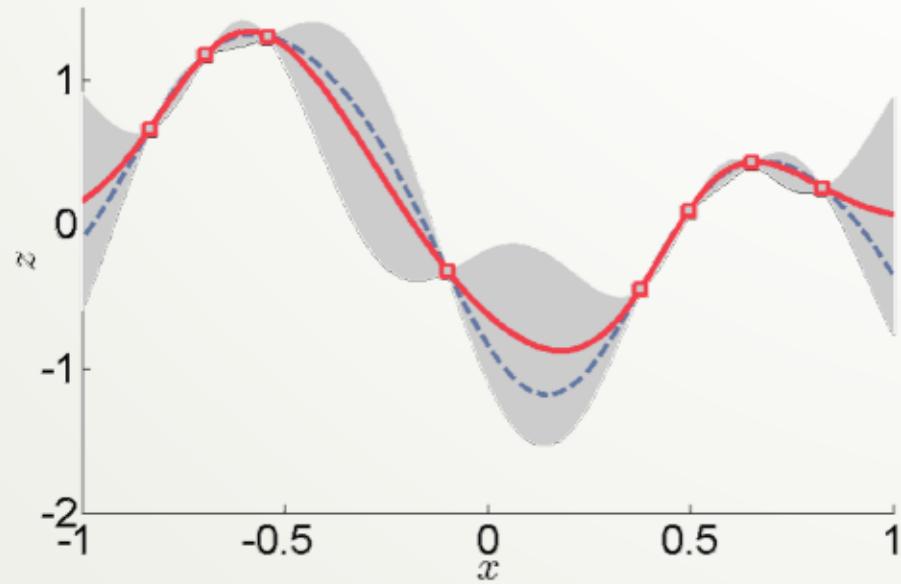
- Open-source
- Vortex Lattice Method (VLM)
- No fuselage
- Wing thickness not taken into account

SU2
code

- Open-source, developed by J.Alonso et al. at Stanford University
- Euler/Navier-Stokes equations

Surrogate models

- What is a surrogate model?
 - Model an unknown function
 - Train from known datapoints (DoE)
 - Create a response surface
 - Predict any new point in the training range

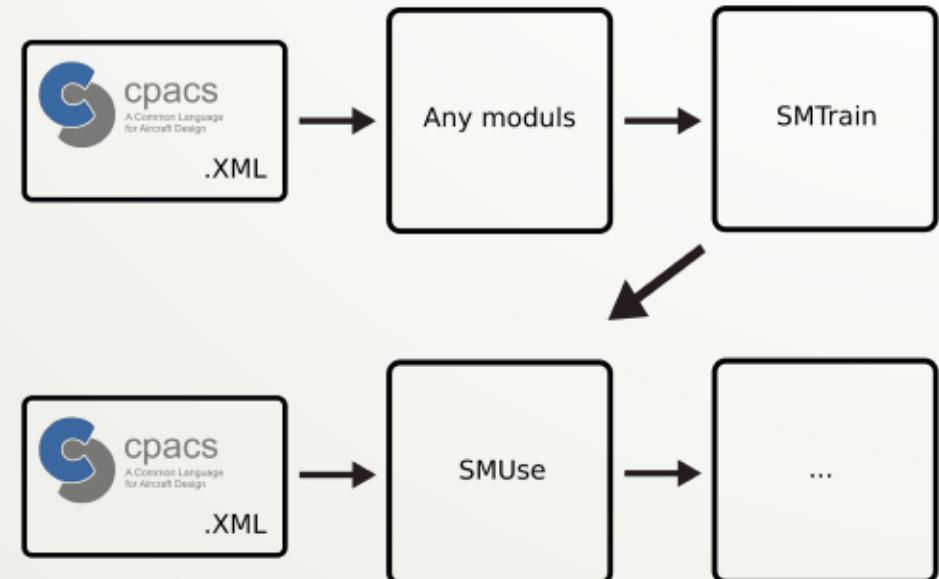


Surrogate models

- Surrogate modeling Toolbox (SMT)
 - Open-source Python package
 - Developed by M. A. Bouhlel et al.
 - Several models implemented
 - Kriging
 - Least square
 - KPLS
 - KPLSK

Surrogate models

- Integration into CEASIOMpy
 - **SMTrain** to train the Surrogate model
 - **SMUse** to predict values using the surrogate model
 - Compatible with all other modules



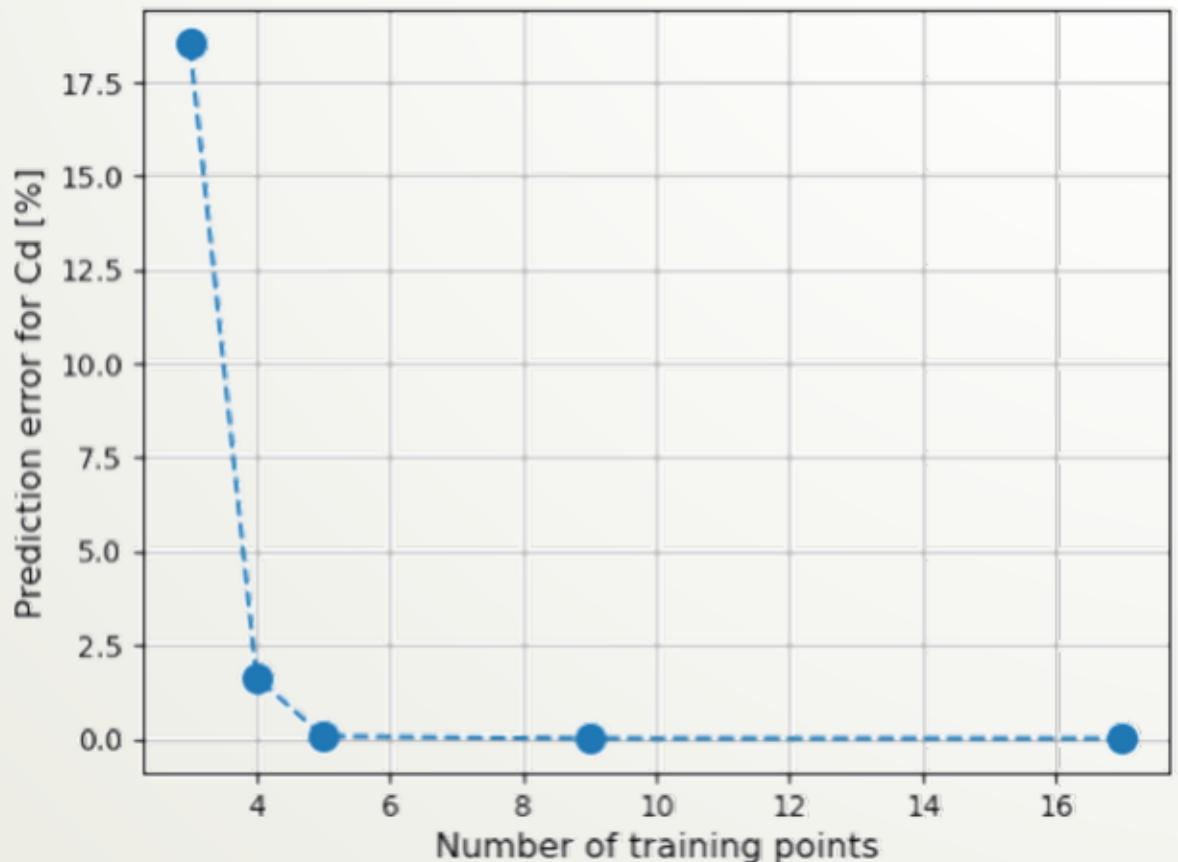
Applications

- Predict aerodynamic coefficients
 - 1 parameter
 - Angle of attack, from -8 to 8 degrees
 - Surrogate model trained with different number of points (17, 9, 5, 4, 3)



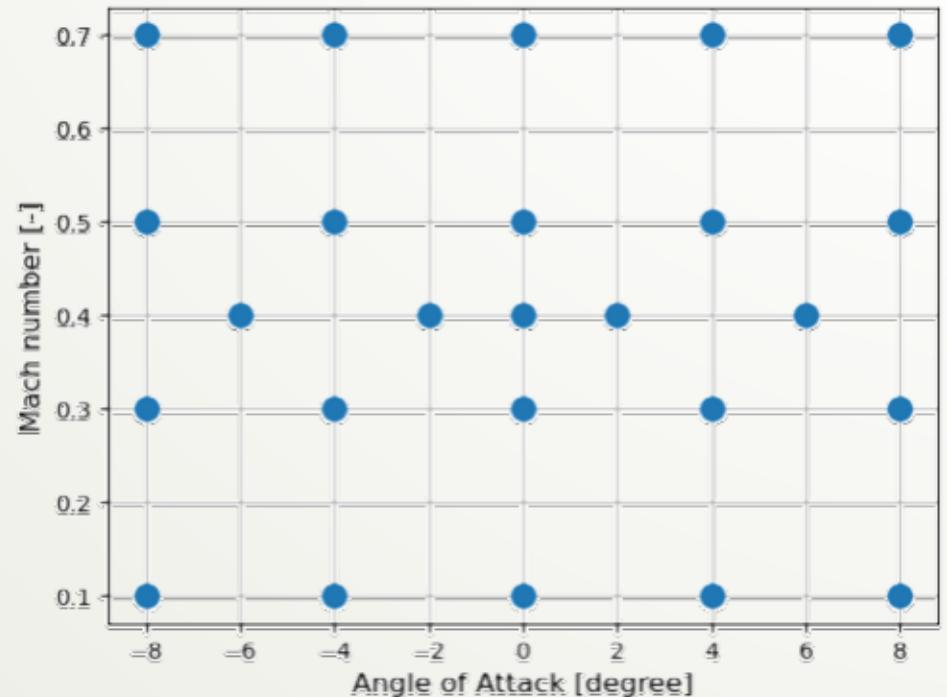
Applications

- Results
 - Error between predicted and calculated Drag coefficient
 - For simple cases 5 points are sufficient



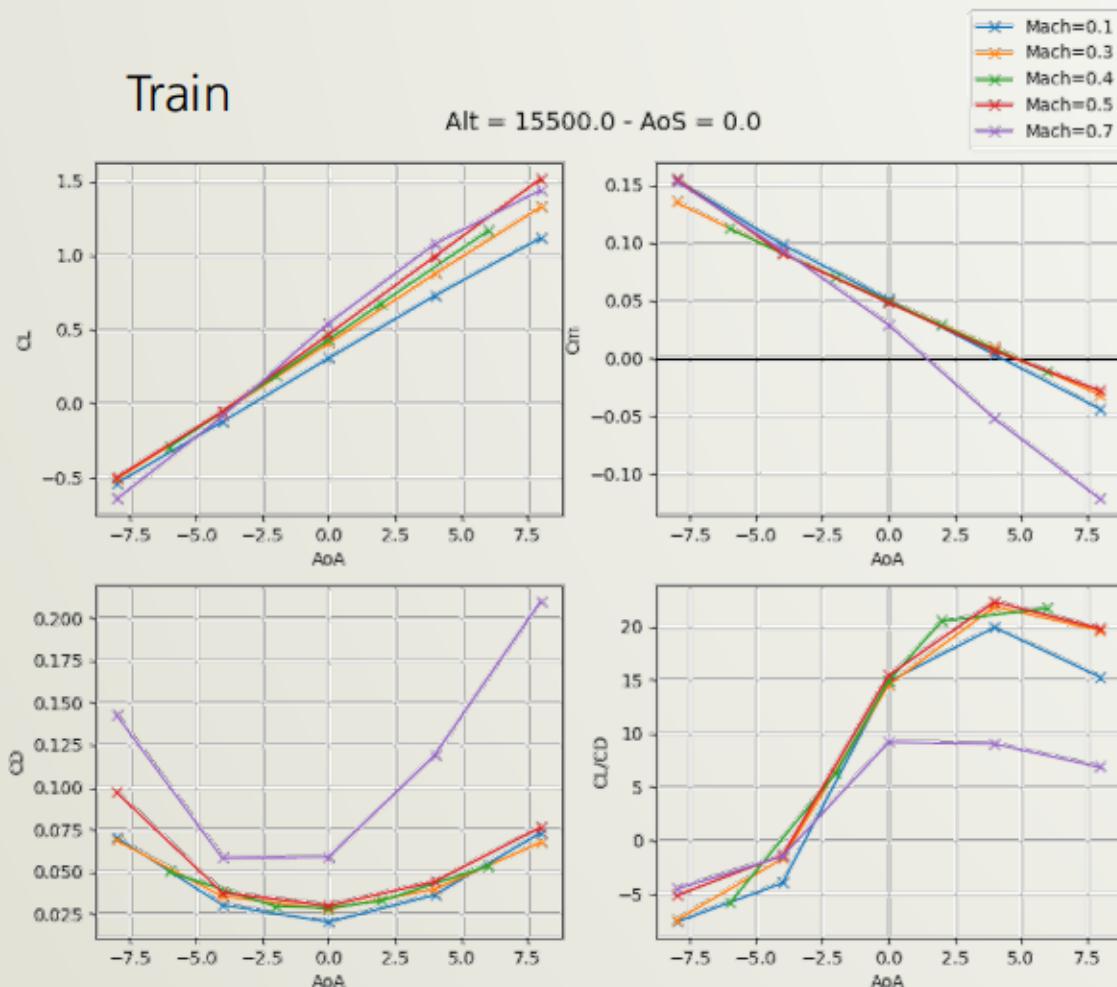
Applications

- Predict aerodynamic coefficients
 - 2 parameters
 - Angle of attack
 - Mach number
 - Training point calculated with:
 - SU2 (Euler) + skin friction
 - Closer to real application

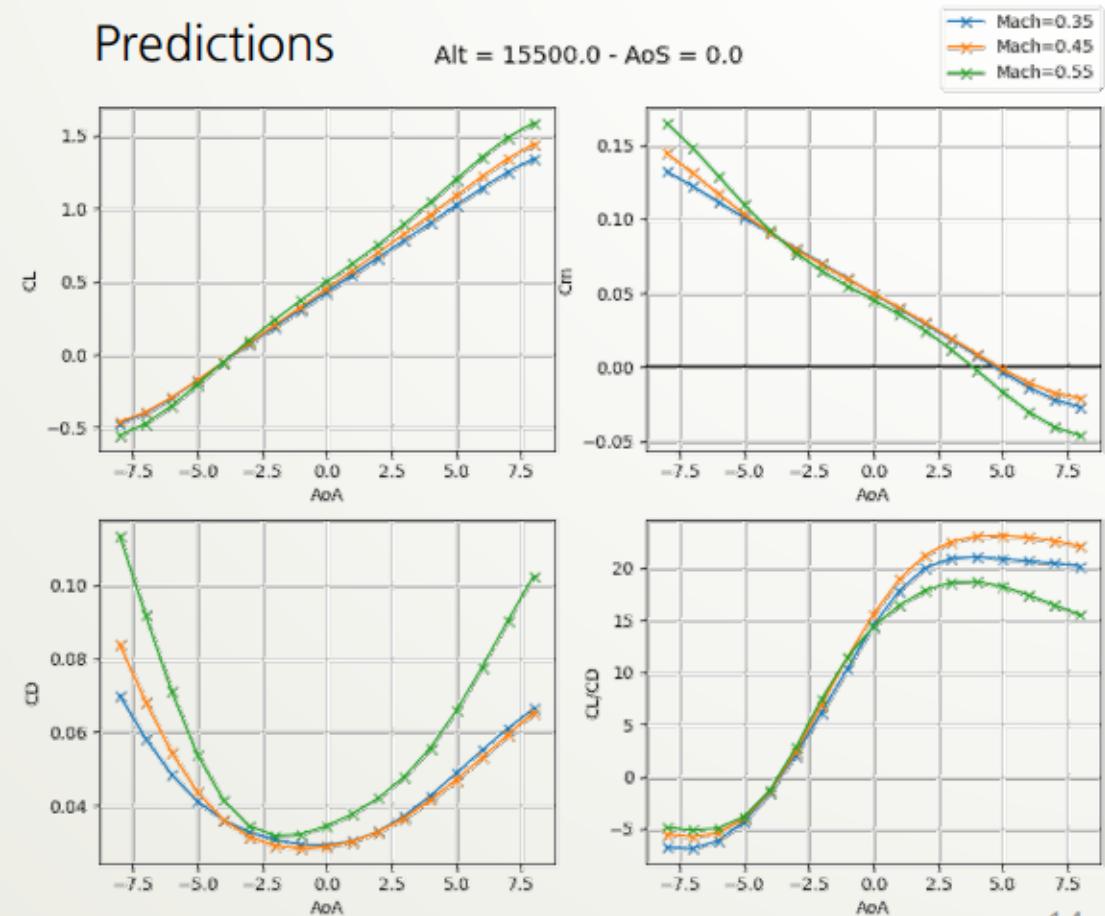


Applications

Train

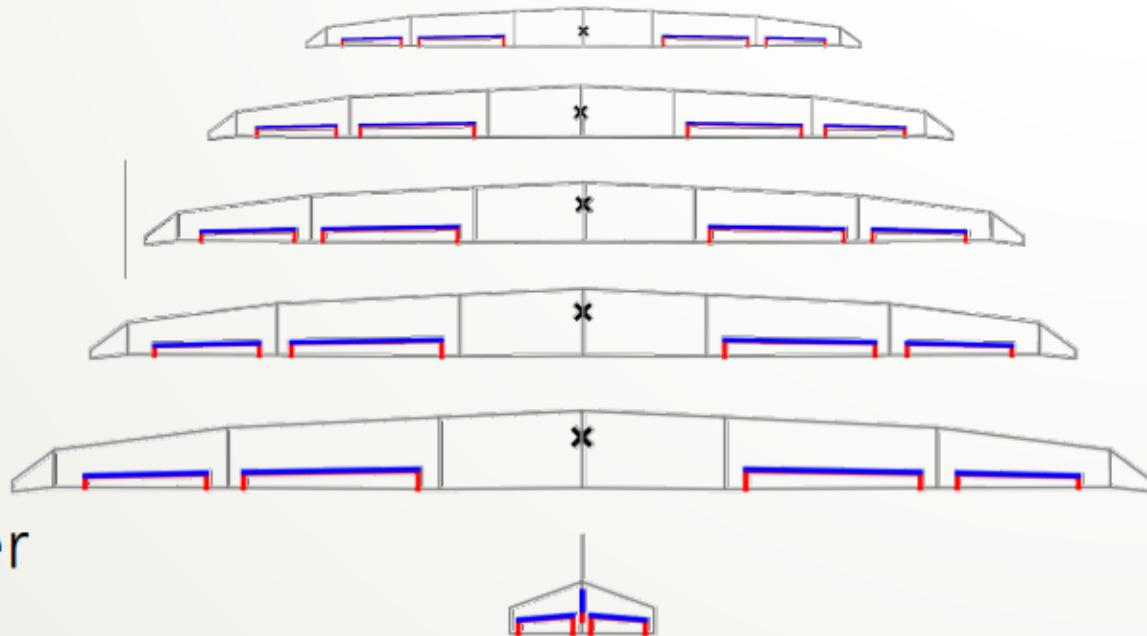


Predictions



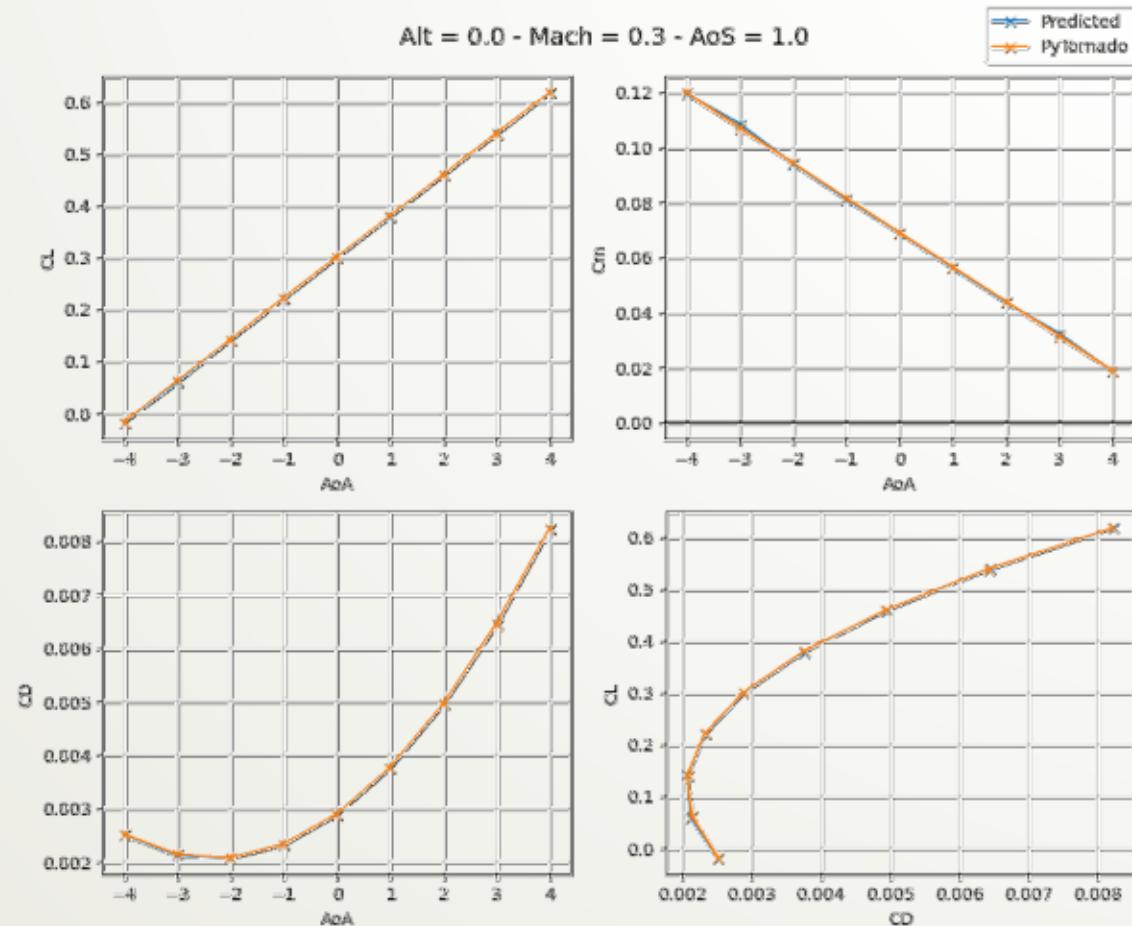
Applications

- Predict aerodynamic coefficients
 - 3 parameters
 - Angle of attack, -4 to 4 degrees
 - Angle of sideslip, -5 to 5 degrees
 - Wing span from, 10 to 20 m
 - Full factorial, 5 values per parameter
 - 125 training points
 - Calculated with PyTornado



Applications

- Results
 - For an uncalculated:
 - Wing span: 14m
 - Angle of sideslip: 1 degree
 - Almost perfect match
 - More points give better results



Conclusion

- Successful integration of SMT into CEASIOMpy
- Many possible applications for
 - Conceptual aircraft design
 - Multi-Disciplinary Optimization
- Room for possible improvement in the integrations
- More tests to find out the best parameters



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