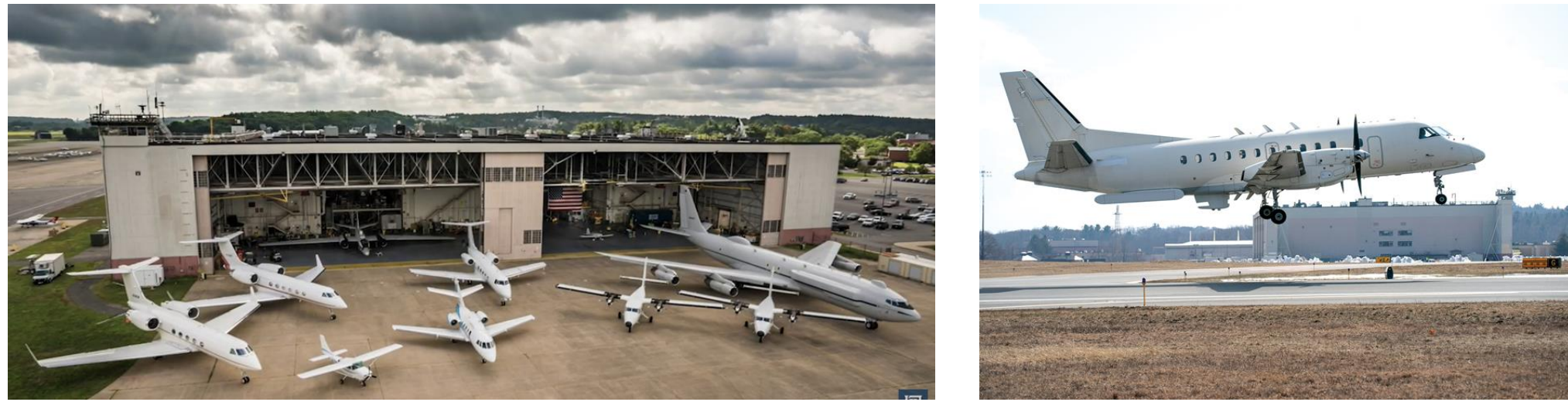
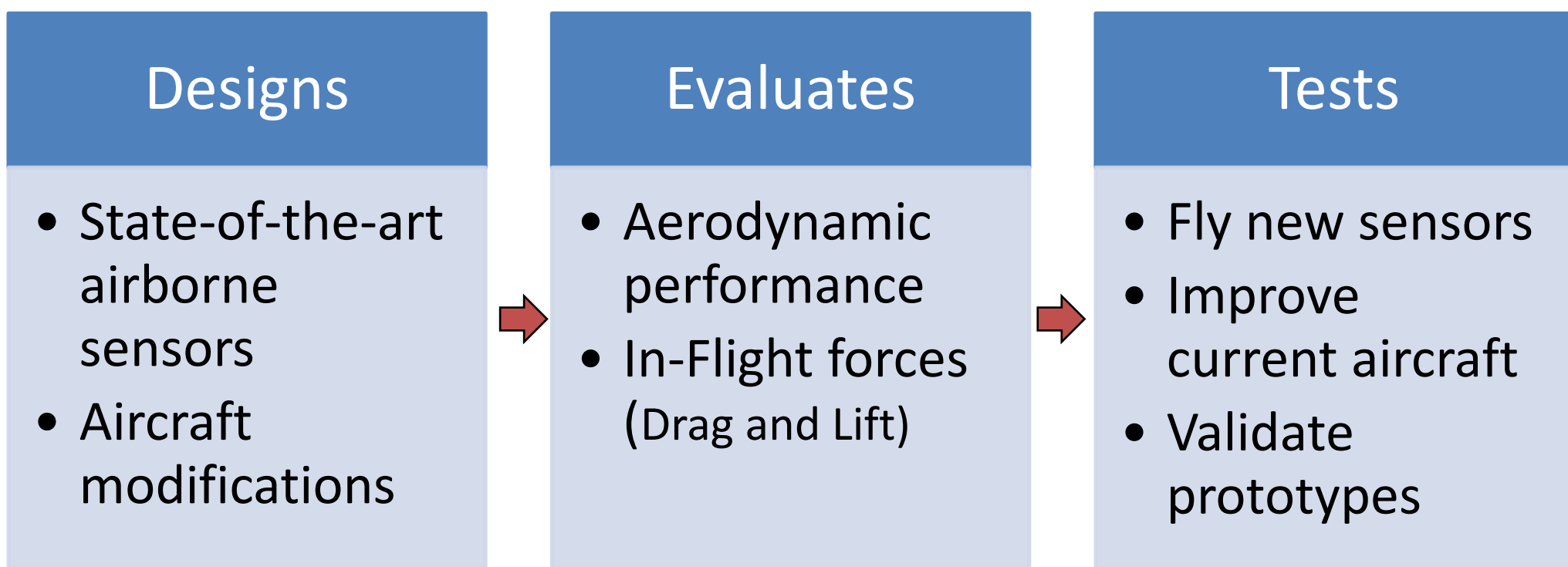


Accelerated Aircraft Prototyping Leveraging CNN's

1. Background: Lincoln Lab's Flight Test Facility

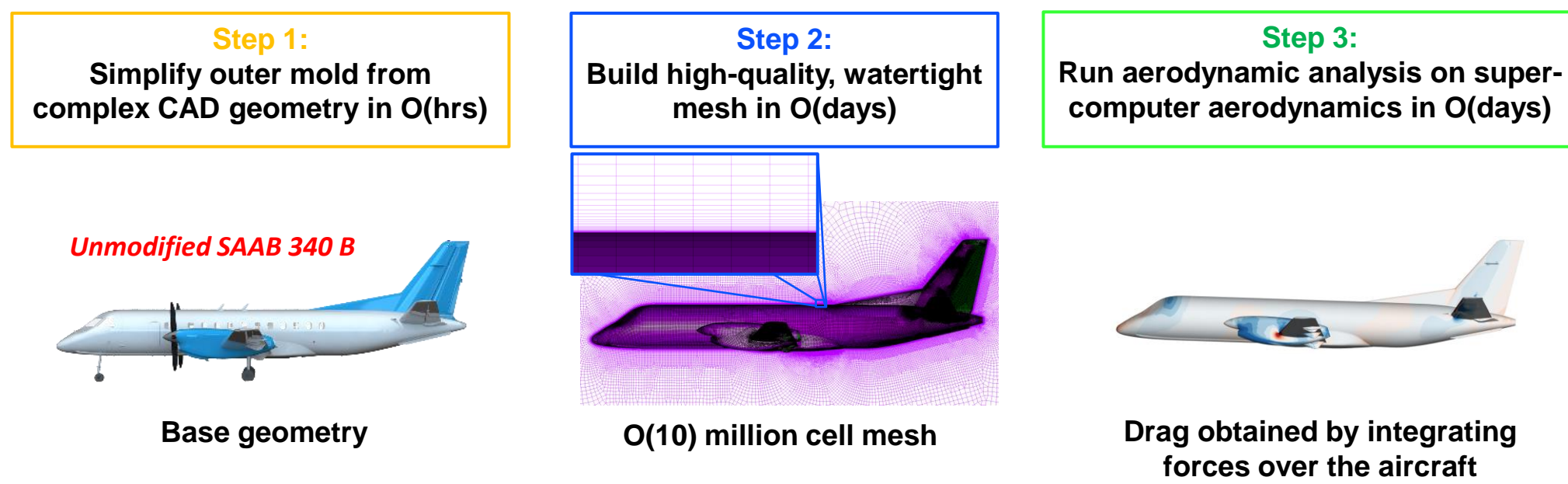


2. Problem: Aerodynamic Analysis is Expensive

Problem Statement

The time and resources required to assess the aerodynamic properties of new aircraft prototypes is a significant hurdle during the conceptual design phase when design parameters are rapidly changing and the opportunity for design impact is highest

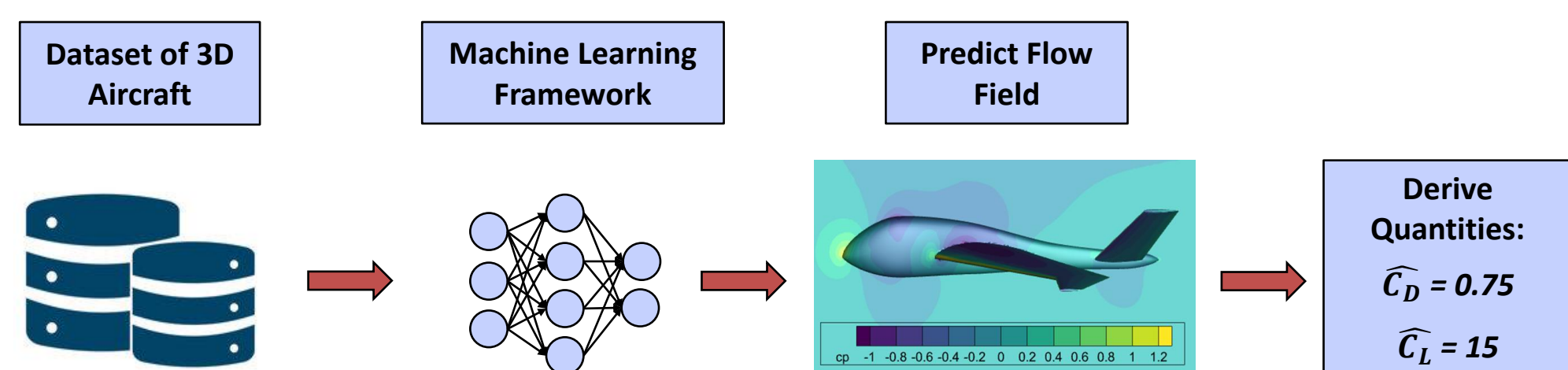
Traditional workflow is labor-, computational-, and time-intensive



3. Goal: Accelerate Aerodynamic Analysis

Objective

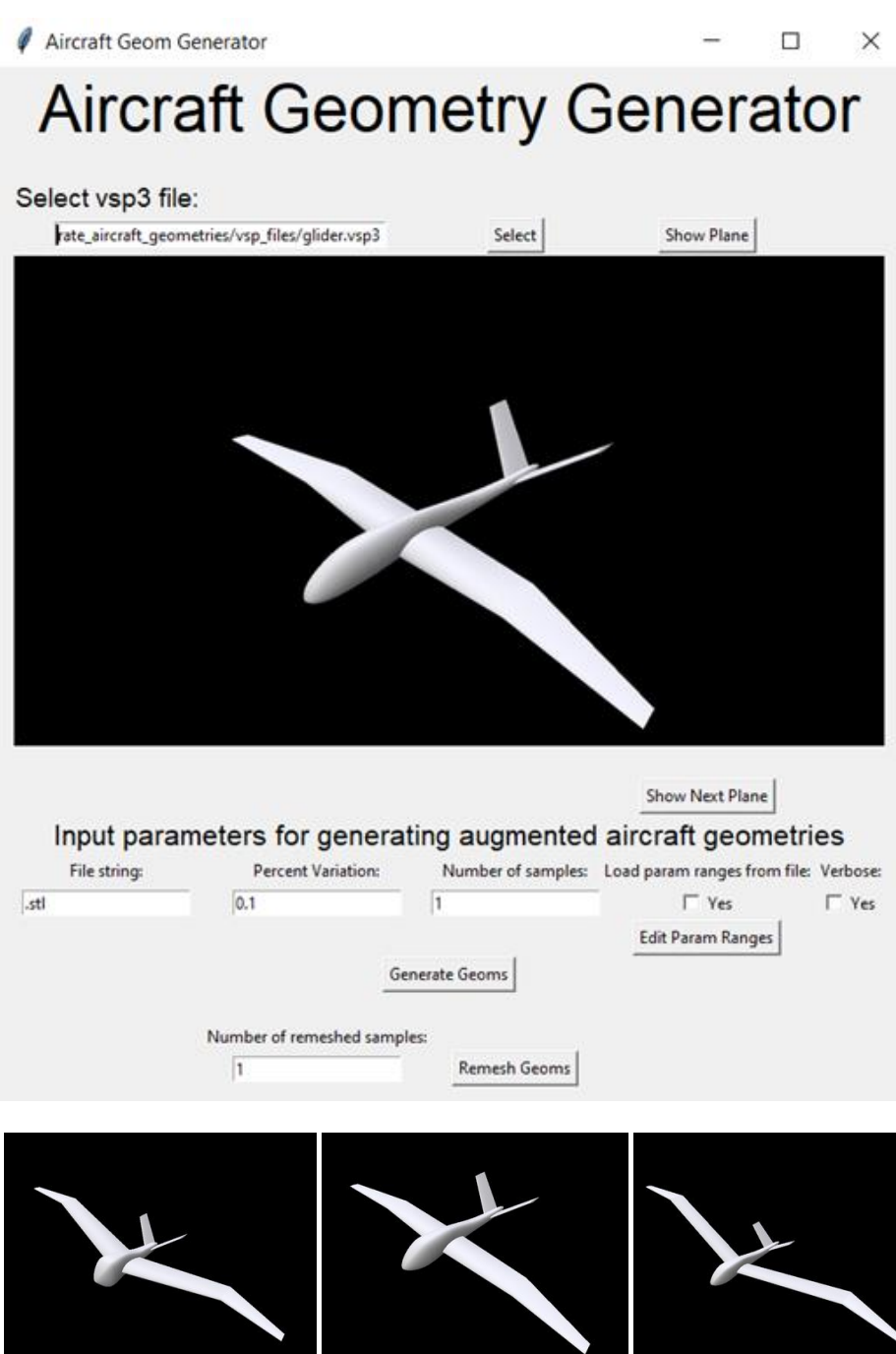
Improve aerodynamic analysis through an automated workflow, leveraging a machine learning framework and expanding the current 3D aircraft model training dataset



4. Solution Part I: Aircraft Data Generation Tool

The Need for a Diverse 3D Aircraft Dataset

- A diverse dataset of aircraft models is necessary to train a generalizable fluids model
- Lincoln Laboratory requires a database representative of most Air Force aircraft
- Certain design parameters are more important to aerospace engineers than others



Aircraft Geometry Generation Tool

- Takes in any aircraft model (OpenVSP)
- Can generate 1000's of new geometries
- Latin hypercube sampling of design parameters
- Can leverage open source community

User Specifications

- GUI and command line functionality
- Input ranges or a percent variation for critical aerodynamic parameters
- Remeshing feature enables flexibility in defining object granularity

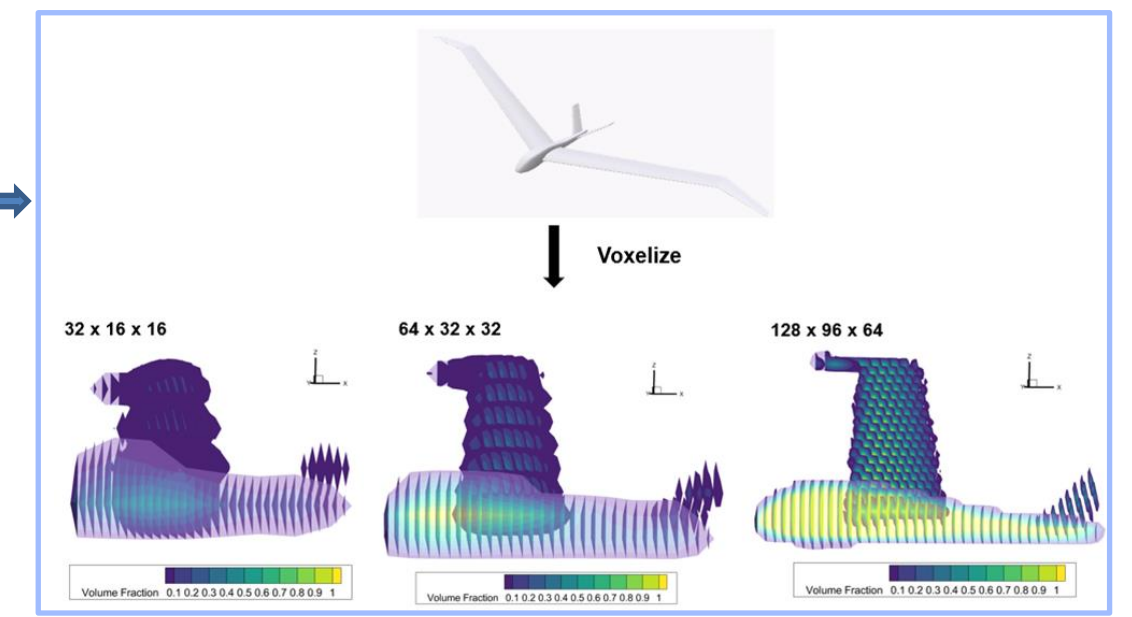
Editable Parameter Ranges

- Fuselage length to average diameter ratio
- Fuselage cross-section shape
- Wing and stabilizer
 - 4. Aspect ratio
 - 5. Taper ratio
 - 6. Airfoil thickness
 - 7. Dihedral angle
 - 8. Twist angle

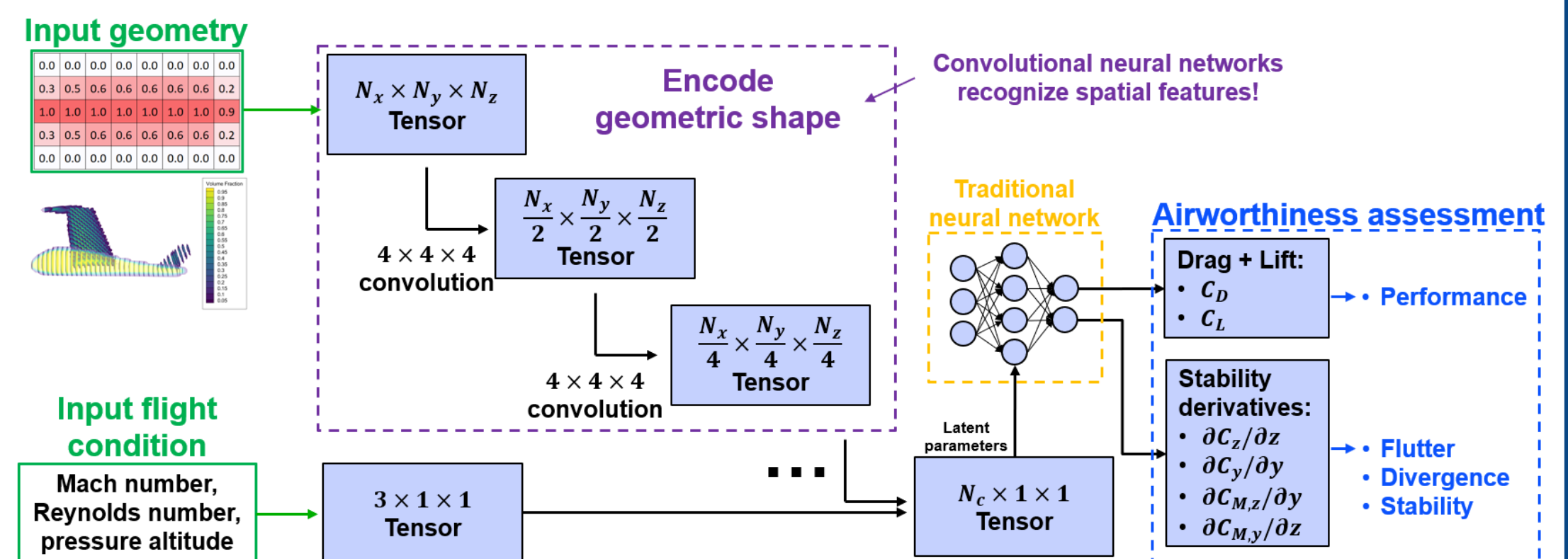
5. Solution Part II: Using Convolutional Neural Networks to Predict Aerodynamic Values

Methodology

- Voxelize geometry into 3D volume tensor
- Different resolutions impact input data
- Concatenate with flight conditions to predict aerodynamic quantities

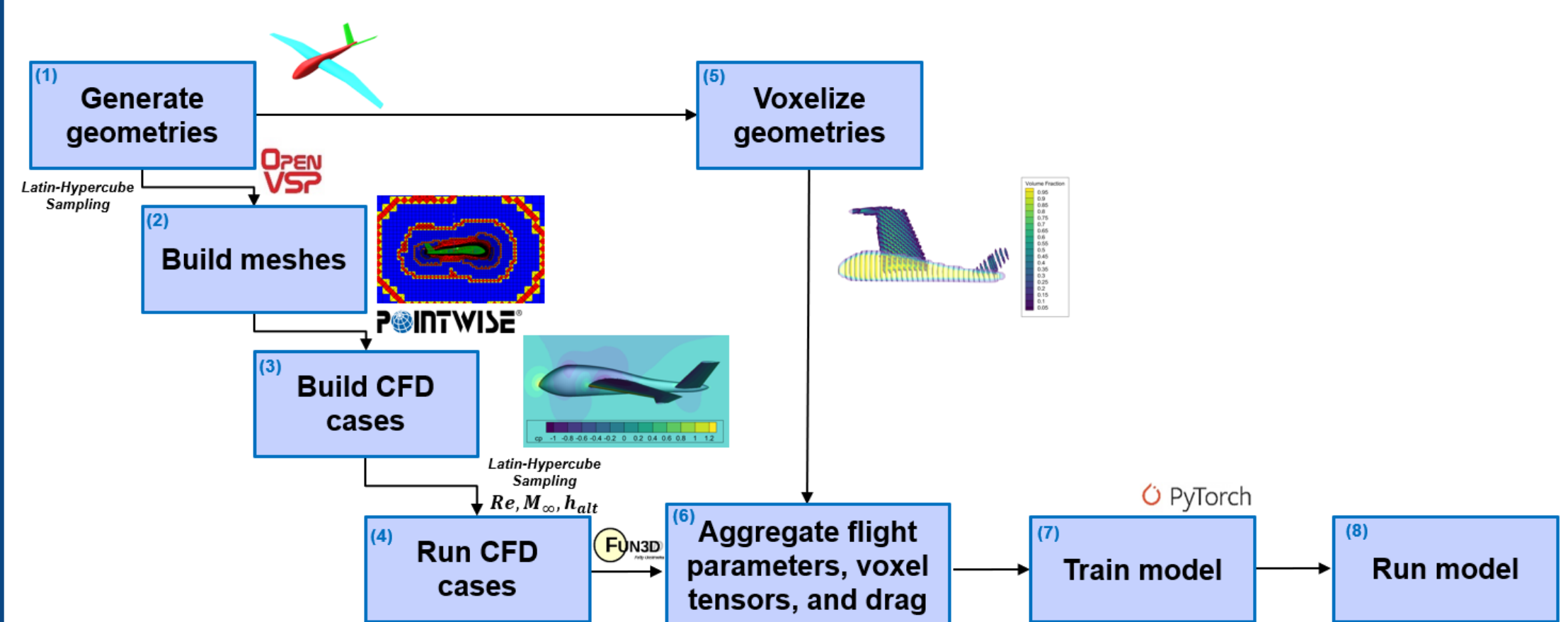


Example Neural Network Architecture



Convolutional Neural Network learns nonlinear relationship between drag for a given geometry and Mach, Reynolds number, and altitude

Training Process Overview

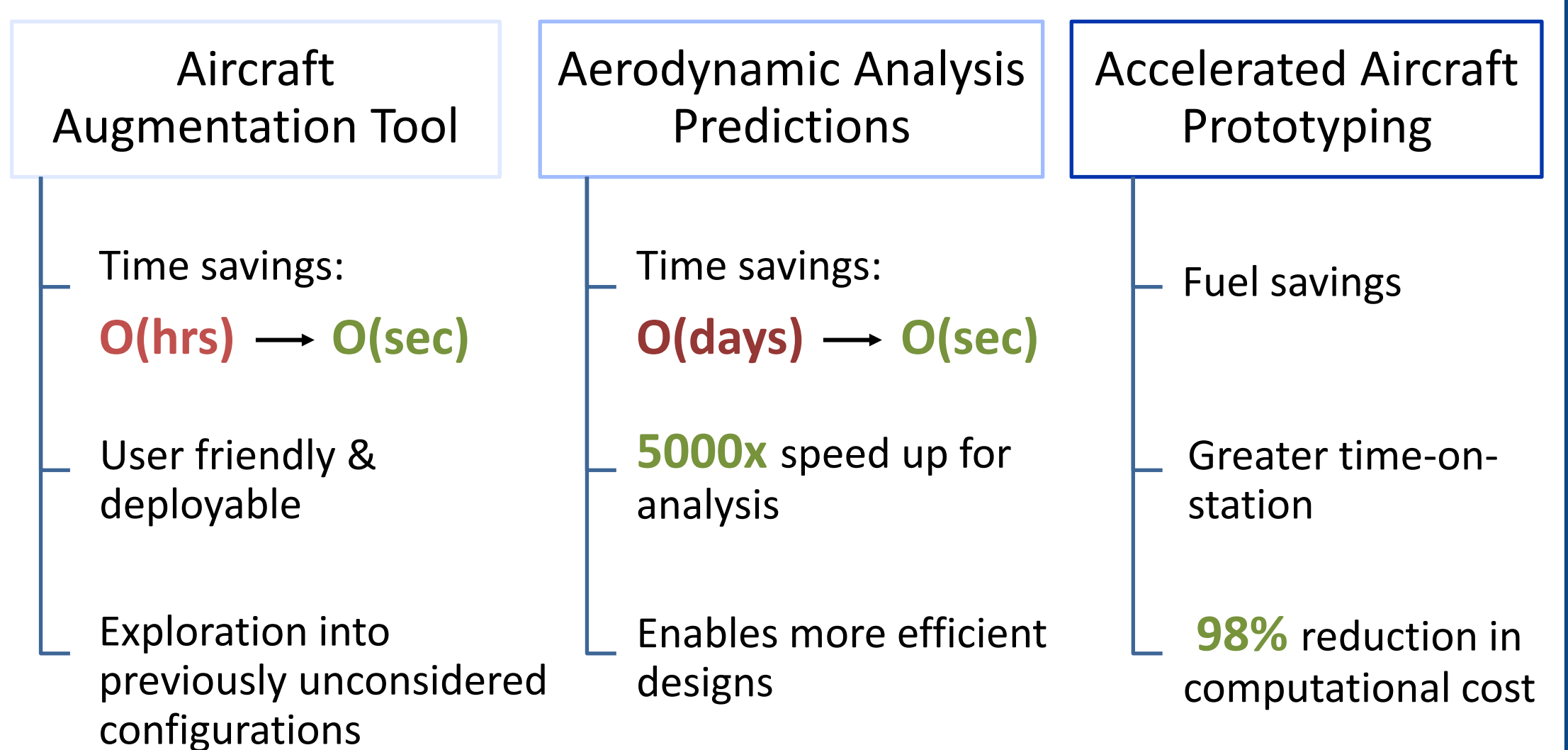


*Ground truth aerodynamic values are obtained from simulations on the GPU accelerated laboratory super computer (90 min each) that require computational meshes (4 min each)

Experiments

- Trained and evaluated models for numerous volume fraction resolutions
- Tested model performance and documented results at different case turbulence tolerances
- Extensive hyper-parameter tuning over various network architectures

6. Impact: Efficient Aerodynamic Analysis



7. Multidisciplinary Innovation

