



Learn-To-Fly Project Overview: ***In-Flight and Wind Tunnel Global Nonlinear Aerodynamic Modeling***

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Outline

- L2F Introduction
- Flight Test Overview
 - Test Aircraft
 - Flight Test Maneuvers
 - Aerodynamic Modeling
 - Example Flight Test Results
 - Real-Time Demo
- Wind Tunnel Test Application
- Summary



Learn-To-Fly Attributes

- **Ultimate Goal: Self-Learning and Autonomously-Adapting Vehicles**
- Potential for High Impact In Three Primary Areas:
 - Safety
 - Reliability
 - Design/Development Efficiency

Technology Enablers Required:

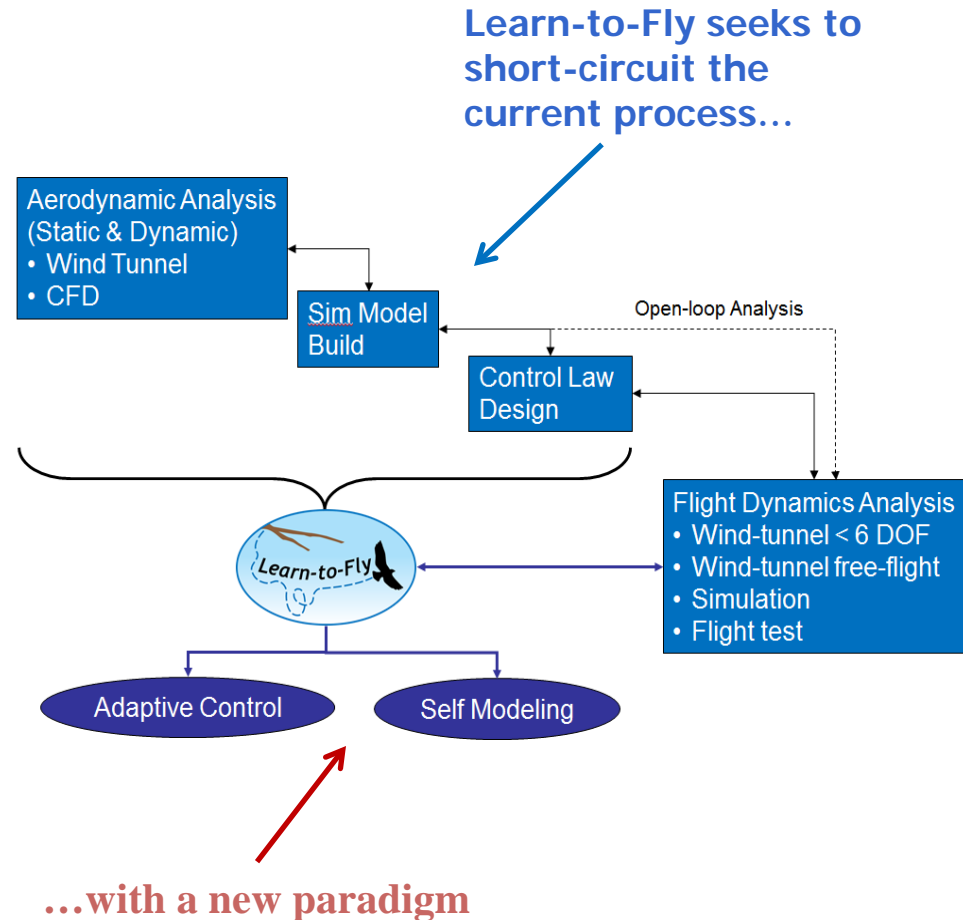
Computer Processing Capabilities

System Identification Advances

- Non-linear aerodynamic modeling
- Real-time system identification

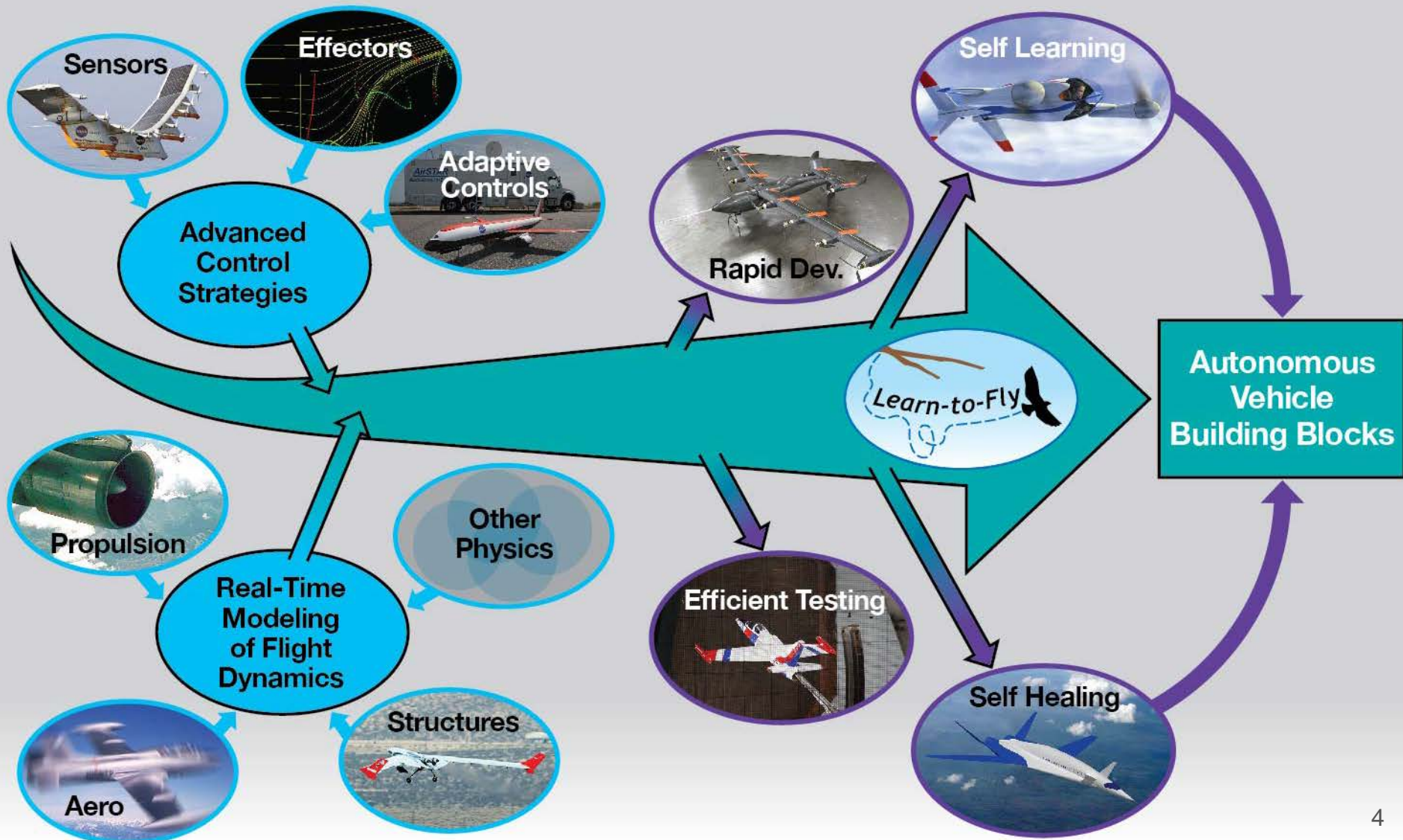
Flight Control Theory Advances

- Adaptive flight control algorithms





Learn-To-Fly Roadmap





Relevant to NASA and LaRC Goals

- NASA Strategic Vision and Langley Goals
 - Transformative
 - On-demand
 - Assured autonomy
 - "Drawings to flight in 30 days"
 - "Learn-To-Fly"
 - Sustainable
 - Intelligent
 - Innovative technologies
 - Global
 - Safety
 - "Refuse to Crash"
 - Efficiency
 - Distributed electric propulsion
 - "Learn-To-Fly" opens design spaces



Recent L2F Results

■ Flight Test

- Supported by Seedling and the National Test Pilot School
- Real-time maneuver and modeling analyses
- Development of near-real-time modeling of 6-DOF nonlinear aero
- Demonstrated in-flight modeling of large envelope, nonlinear aerodynamics

■ Wind Tunnel Based Nonlinear Modeling

- Supported by the Aeronautical Sciences Project - Advanced Aircraft Flight Controls
- Developmental study using wind tunnel testing
- R/C L-59 model tested in 12-Foot Low-Speed Tunnel
- Development of technology and modeling techniques
- Inexpensive & quick



Flight Project Overview

■ Competitively Selected Project

- Sponsor: NASA Aeronautics Research Institute
- 320 proposals from 8 NASA Centers, 20 selected for Phase I
- Phase II awarded
- Final set of flights conducted in February 2014



■ Cooperative Project With NTPS

- Co-PI Gene Morelli





Flight Test Aircraft

Aermacchi Impala MB-326M



- Large flight envelope, aerobatic capability
- Well instrumented:
 - α/β on nose boom, dynamic pressure, engine rpm, fuel flow
 - accelerations, angular rates, and controls
 - real-time data in aft cockpit and telemetered to the ground



Flight Test Setup

- Standard Front Cockpit
- Onboard Instrumentation
- Real-time Analysis Computer mounted in Aft Cockpit





Data Input Maneuvers

■ Conventional Approaches

- Trimmed flight – time consuming single point conditions
- Normal flight regime
- Single axis at a time
- Stitched together model

Correlation Coefficient

$$\rho_{u_1 u_2} = \frac{\sum_{i=1}^N [u_1(i) - \bar{u}_1][u_2(i) - \bar{u}_2]}{\sqrt{\sum_{i=1}^N [u_1(i) - \bar{u}_1]^2} \sqrt{\sum_{i=1}^N [u_2(i) - \bar{u}_2]^2}}$$

■ Global Modeling Maneuvers

- Optimized multi-axis automated inputs
- Fuzzy piloted inputs
- Fuzzy inputs with changing flight conditions
 - Decel
 - WUT/WDT
 - SPAZ
 - Spin / departures
- Single global model



Comparison of Input Type



Conventional Doublets

- Trim
- Sequential inputs

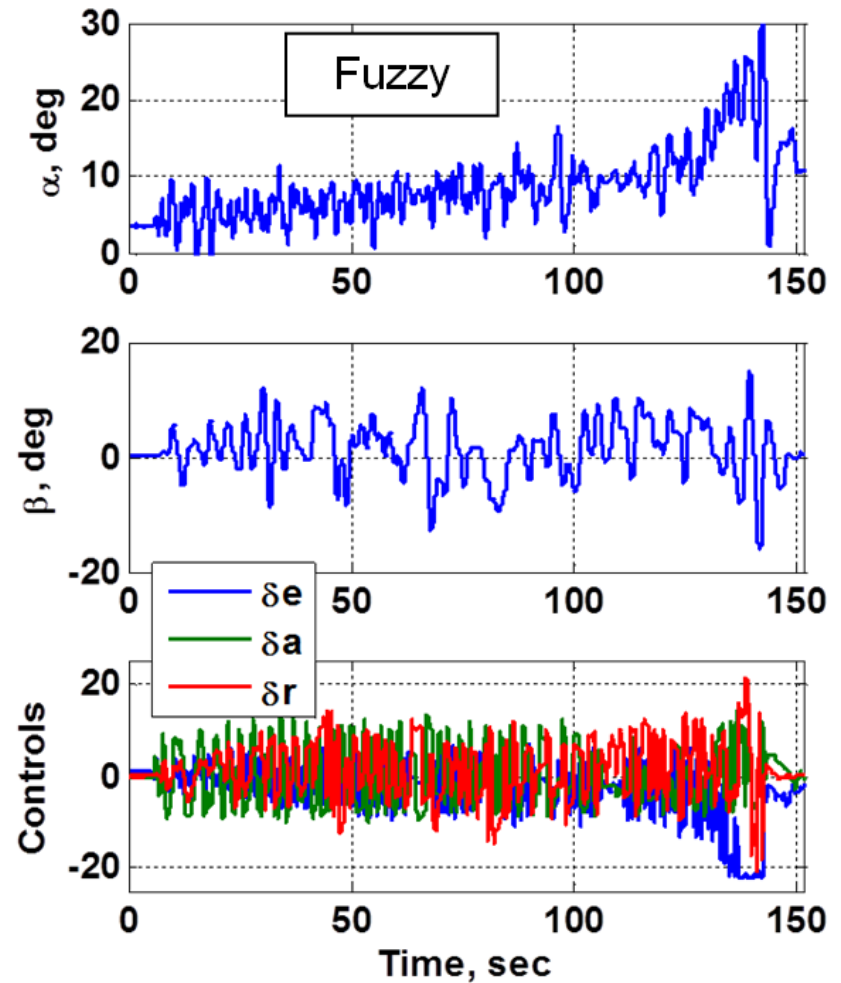
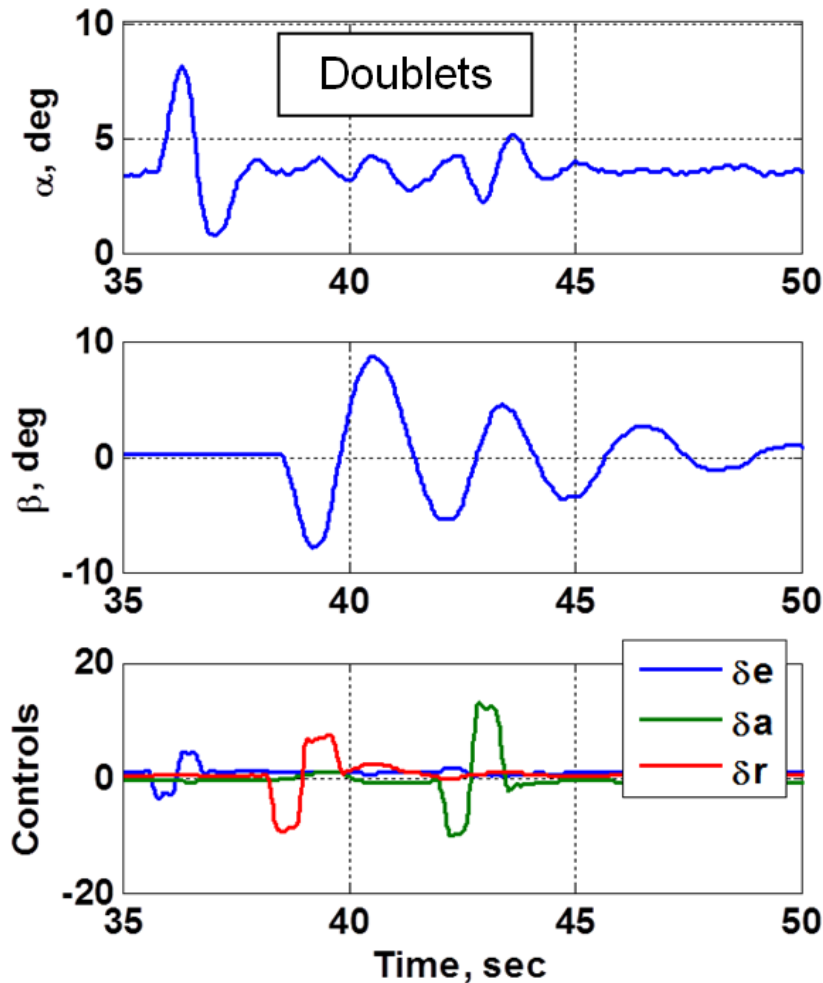


Fuzzy Inputs

- Trim
- Simultaneous uncorrelated inputs

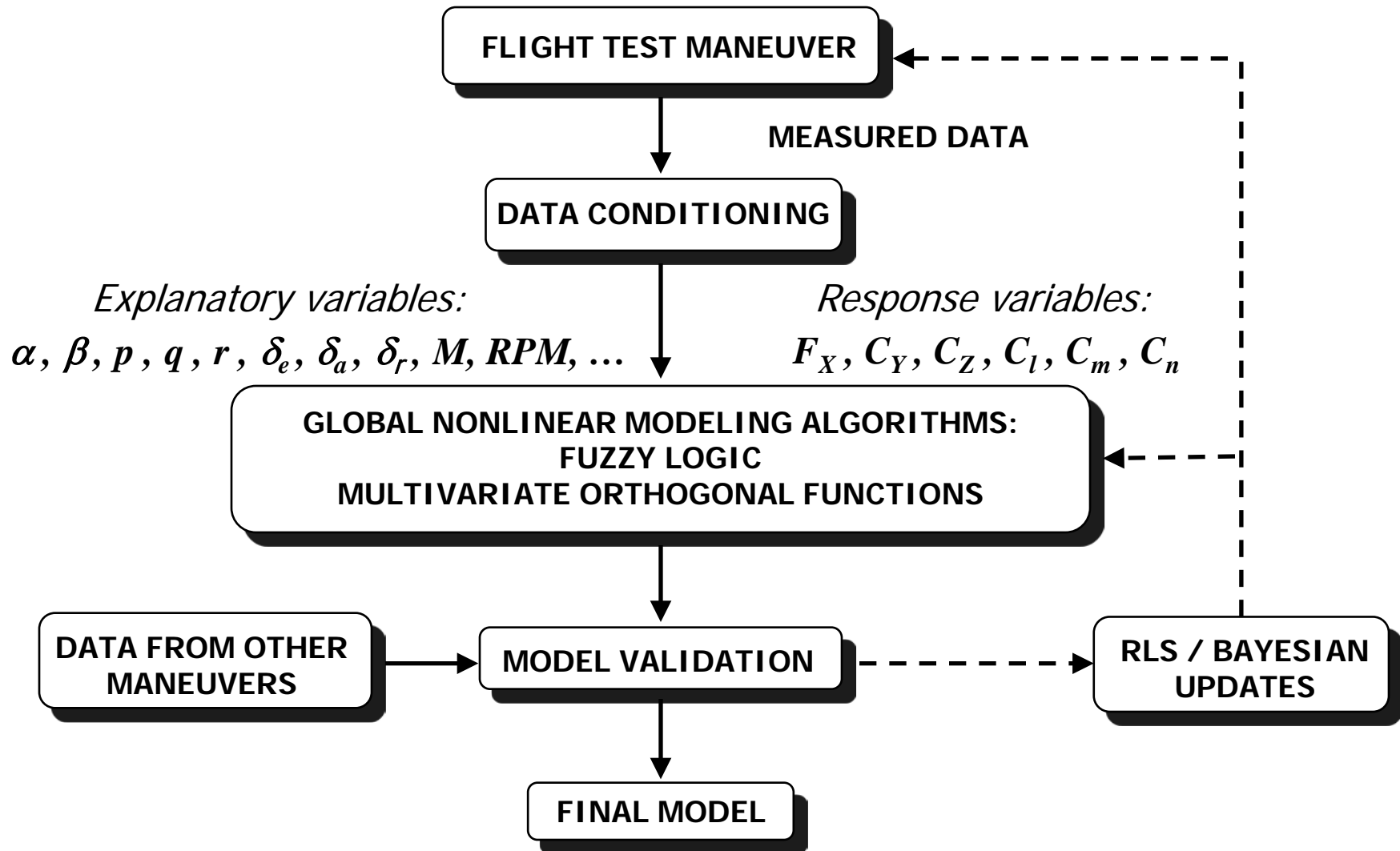


Flight Test Maneuvers



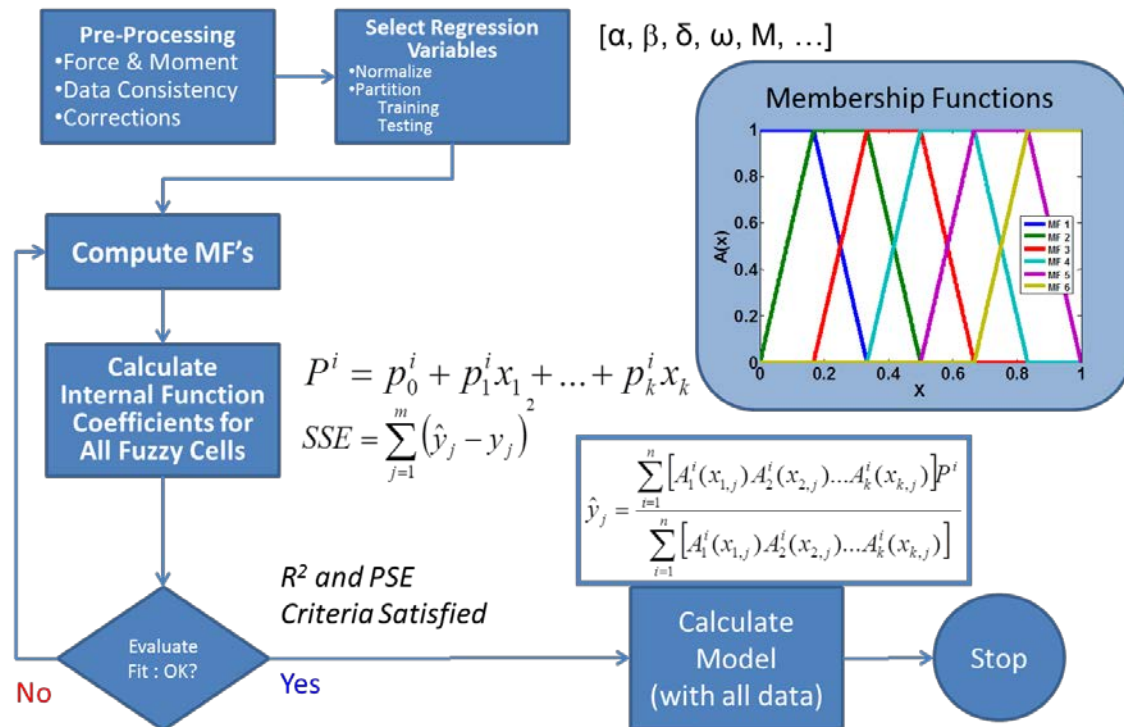


In-Flight Aerodynamic Modeling



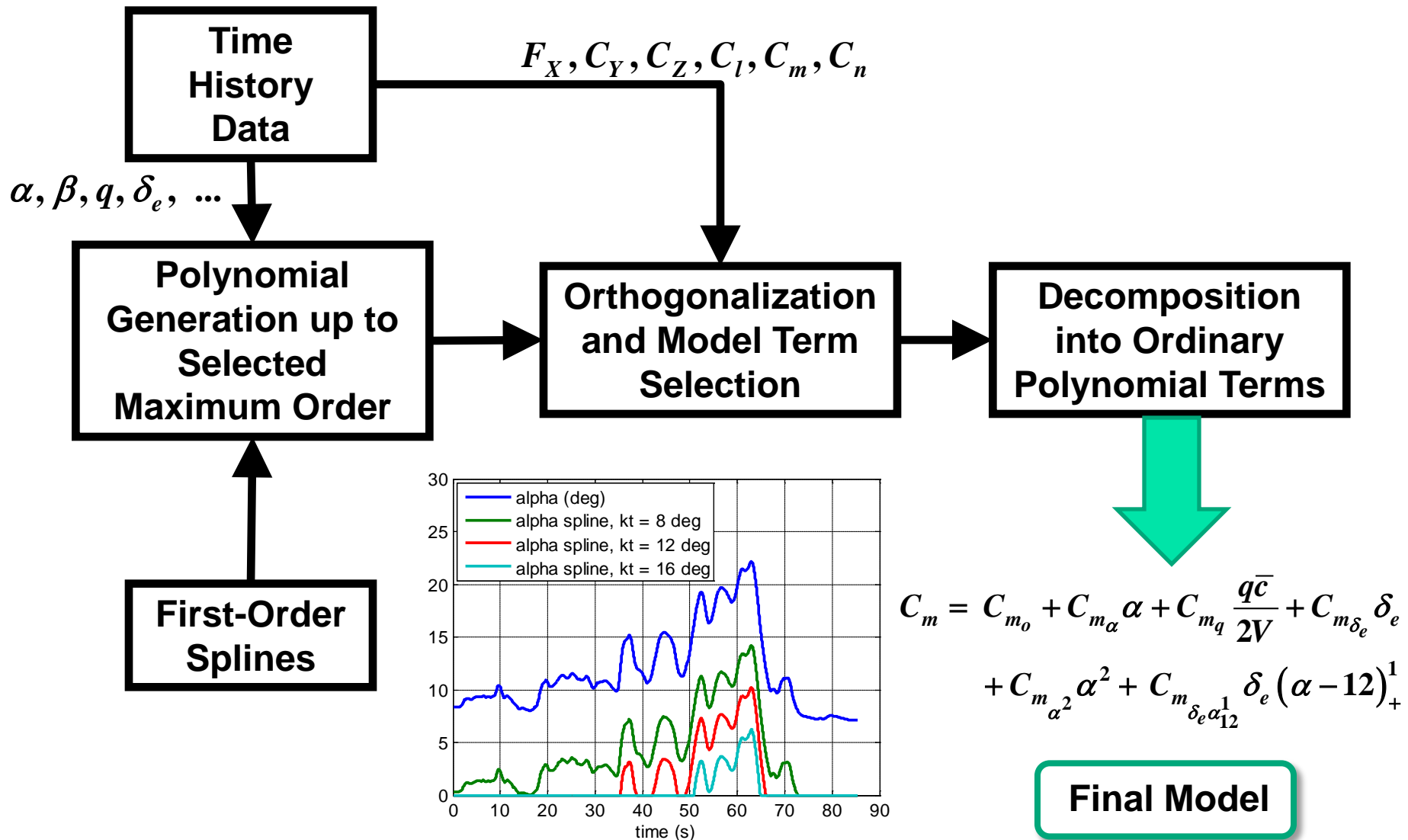
Fuzzy Logic Modeling

- No a priori aero model construct
- Fuzzy cells constructed to identify relationships between input and output data
- Multiple internal functions create the “fuzziness”
- Single model across wide range of state variables
- Predicted outputs are smooth





Multivariate Orthogonal Function Modeling with Splines

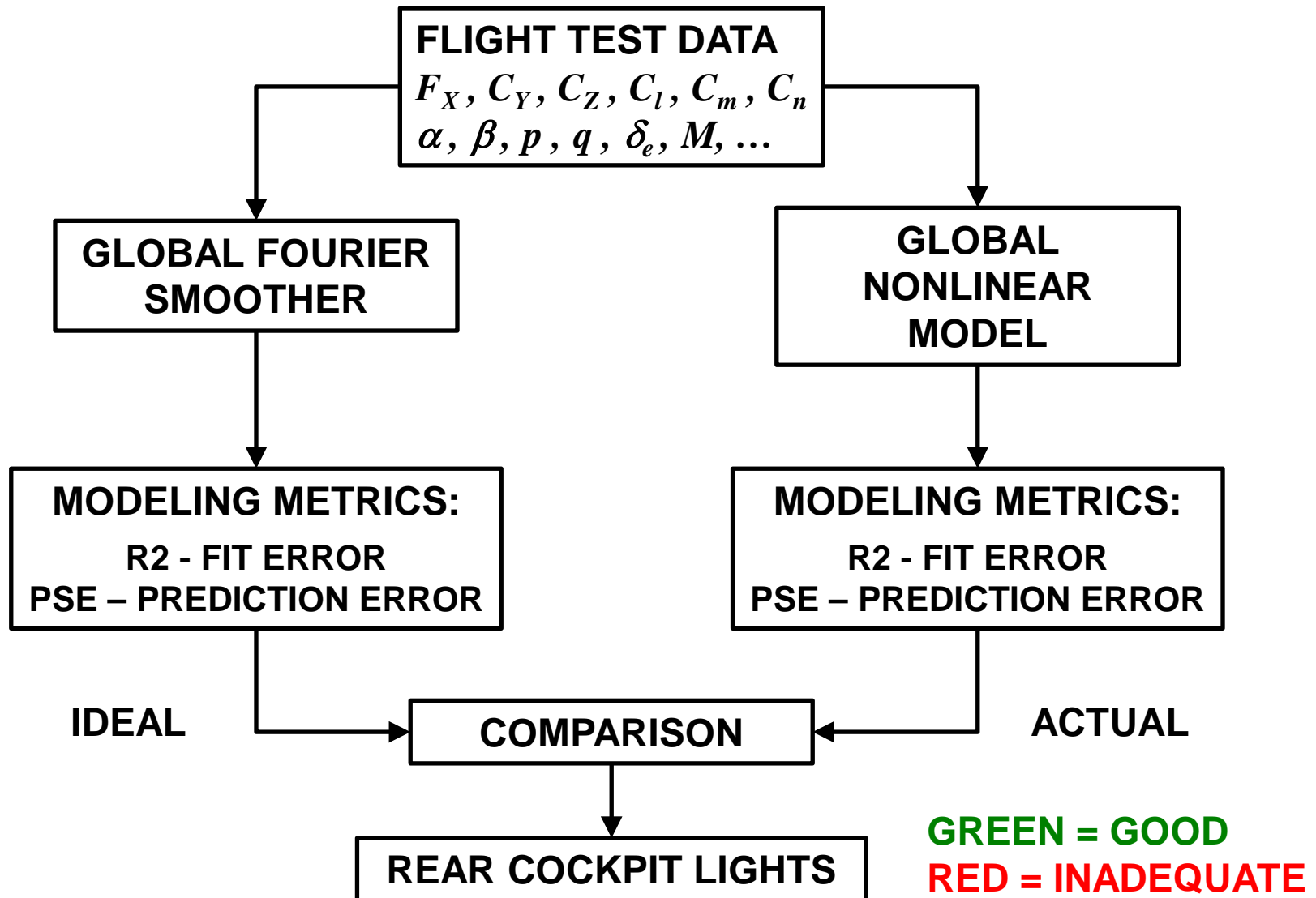


$$C_m = C_{m_0} + C_{m_\alpha} \alpha + C_{m_q} \frac{q\bar{c}}{2V} + C_{m_{\delta_e}} \delta_e + C_{m_{\alpha^2}} \alpha^2 + C_{m_{\delta_e \alpha_{12}^1}} \delta_e (\alpha - 12)_+^1$$

Final Model



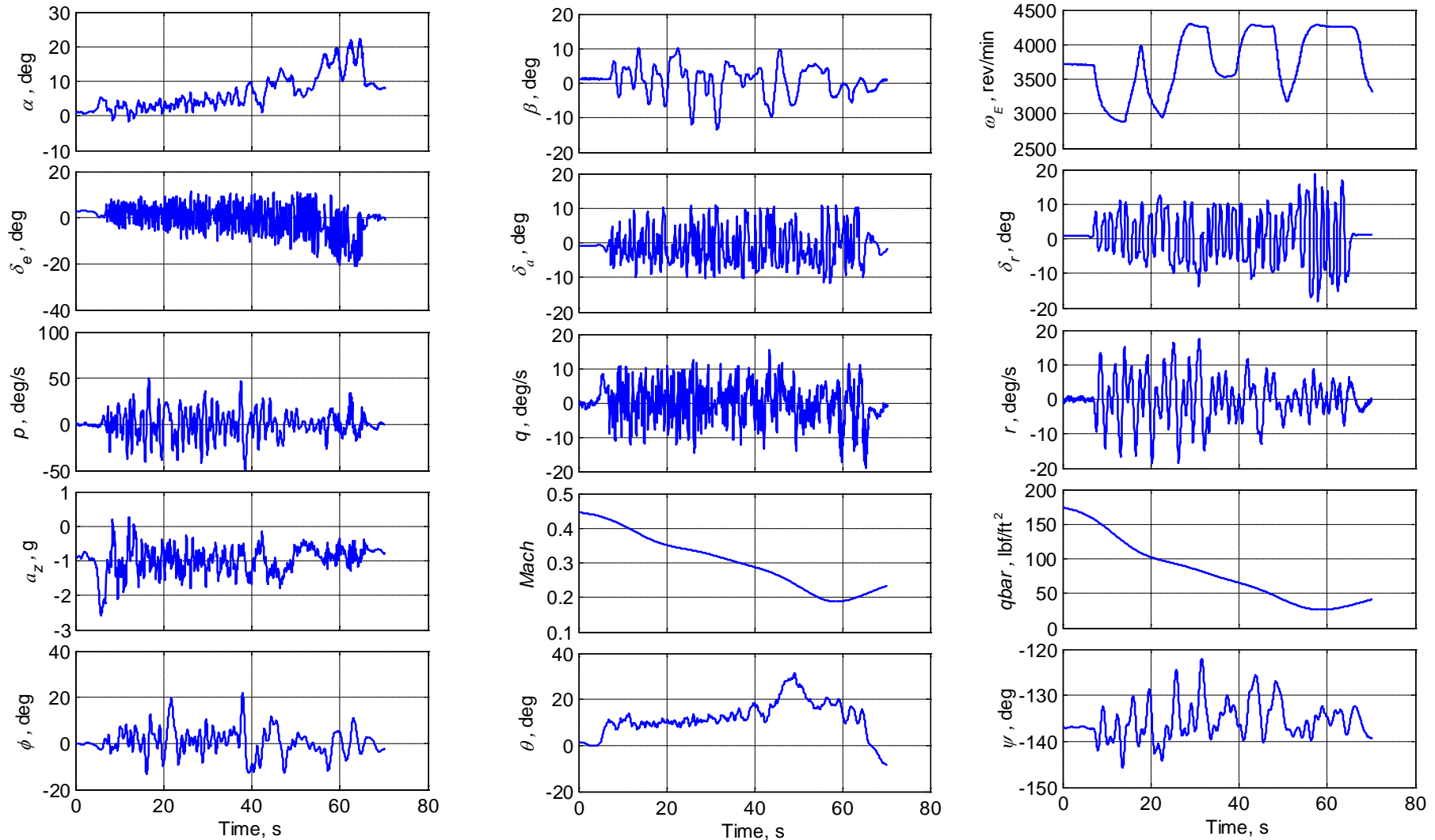
In-Flight Model Evaluation





Powered Fuzzy Deceleration

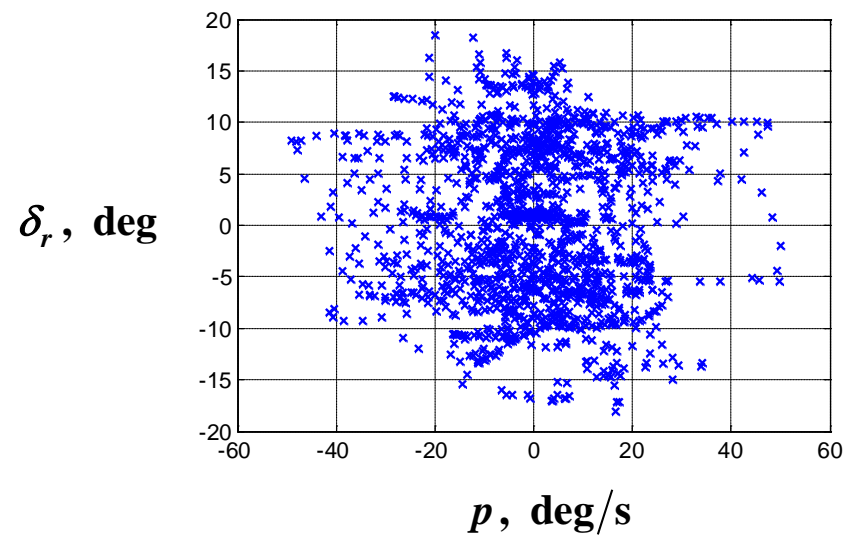
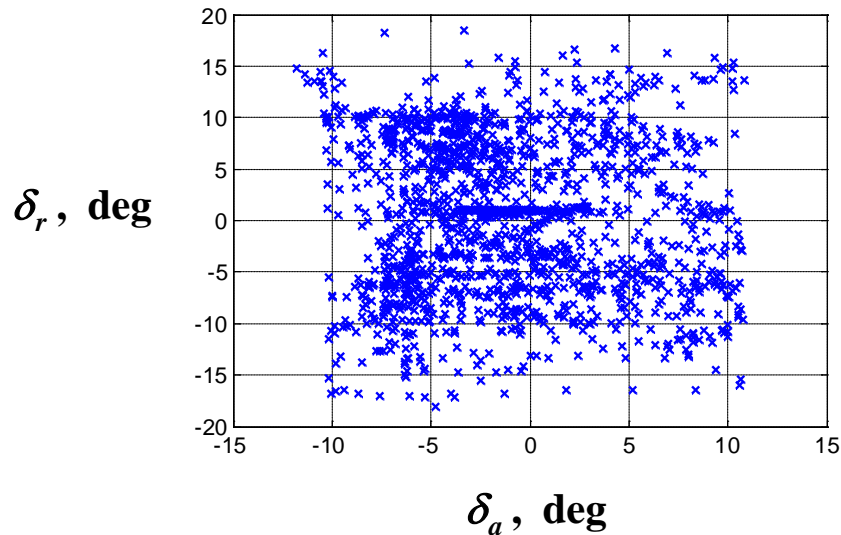
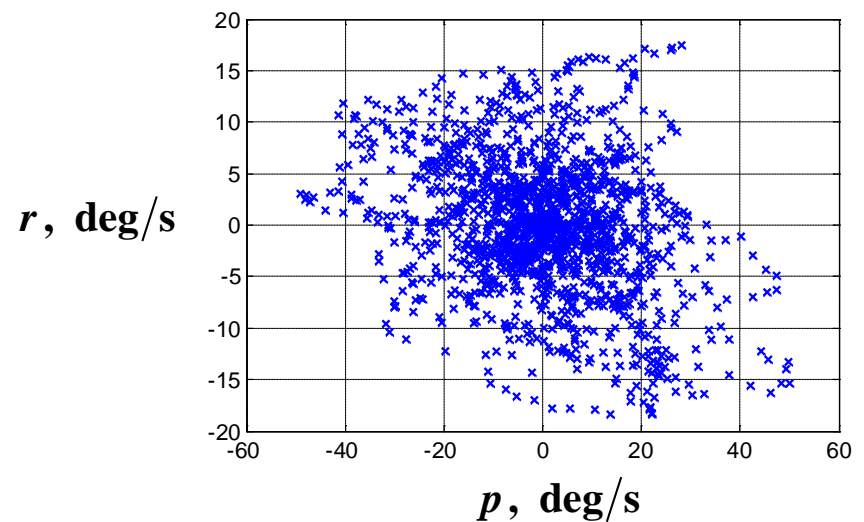
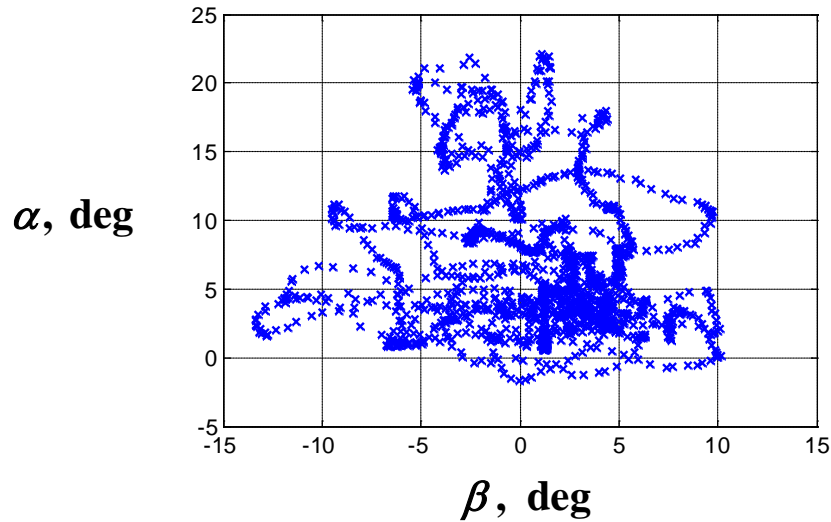
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Explanatory Variable Coverage

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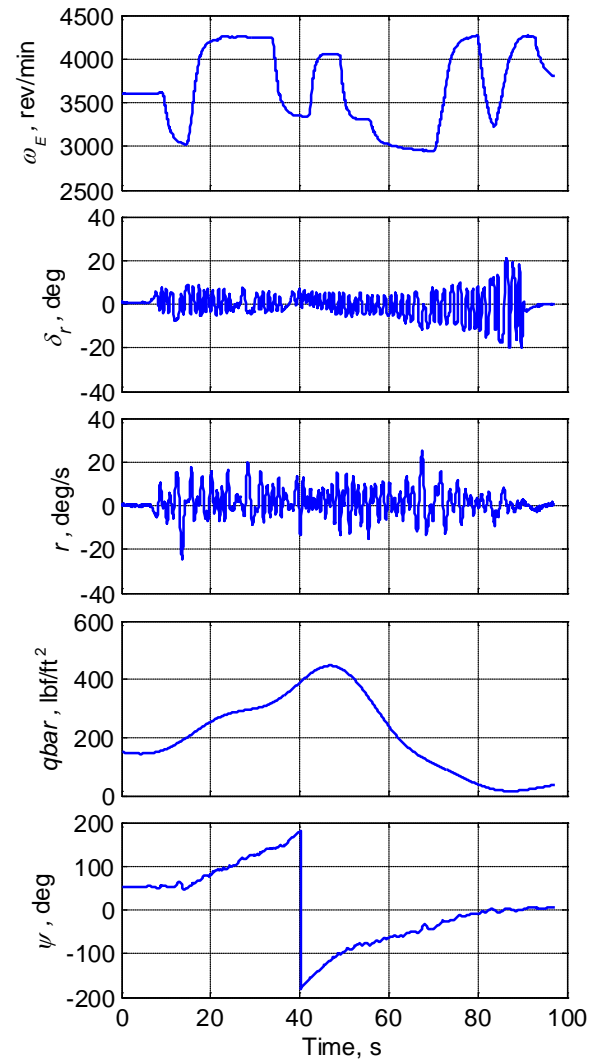
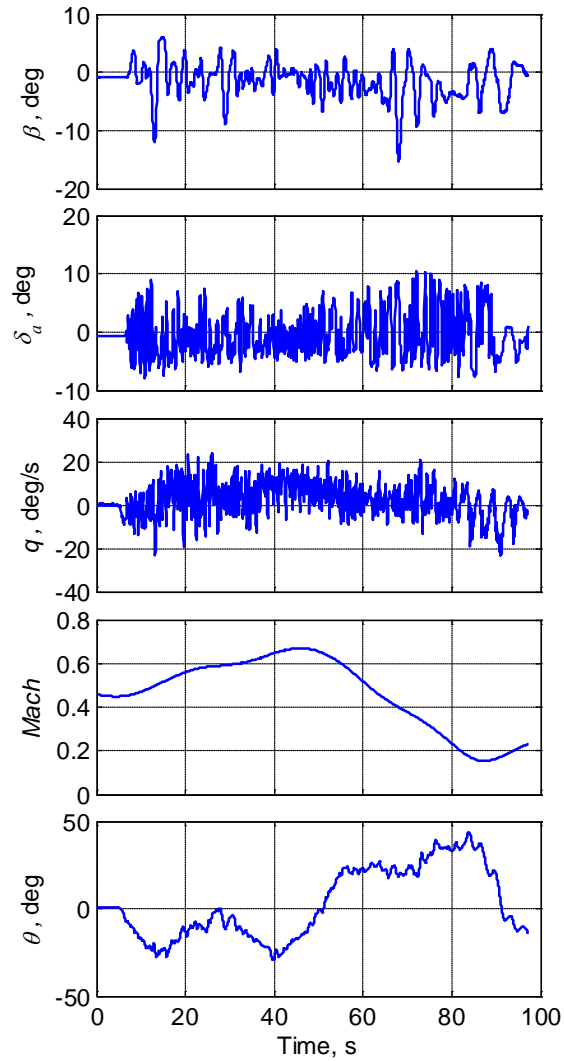
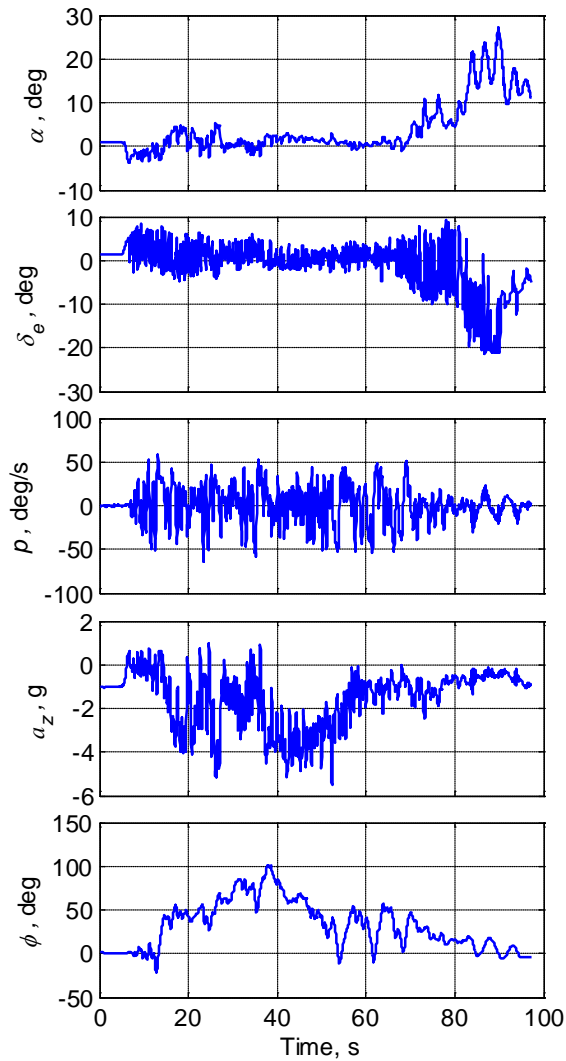
Video of SPAZ Maneuver





SPAZ Maneuver

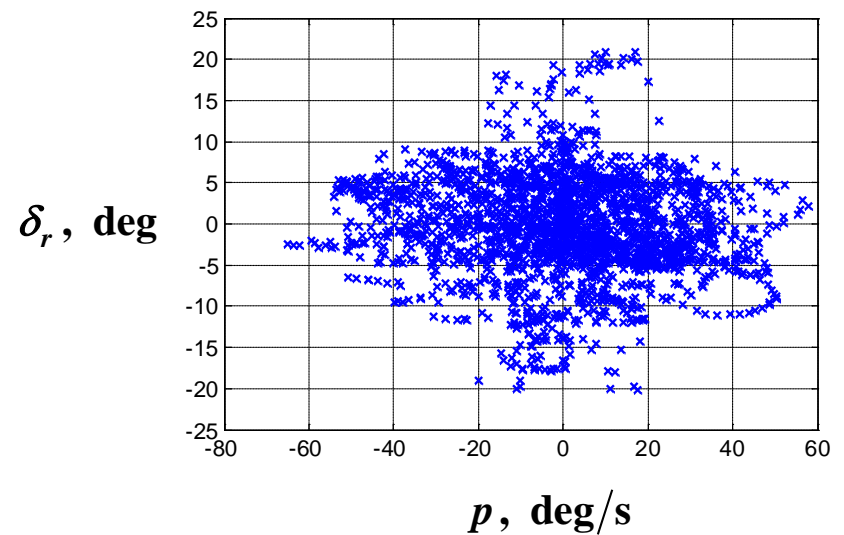
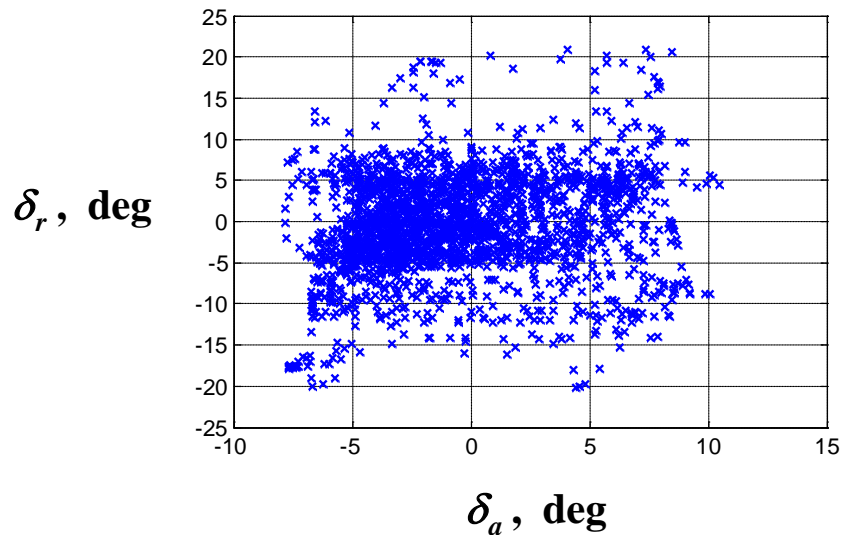
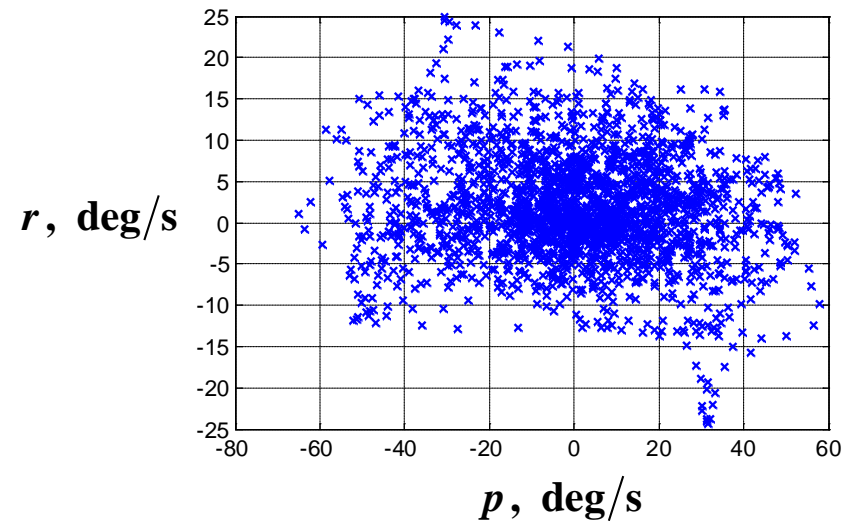
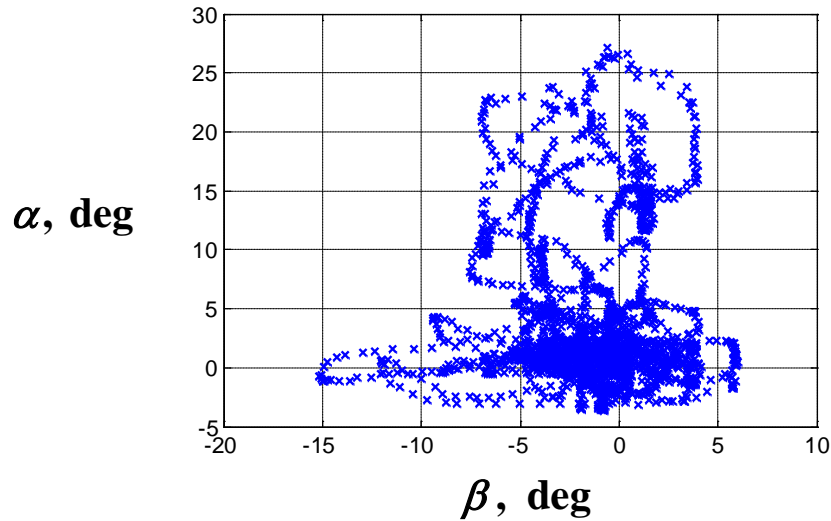
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Explanatory Variable Coverage

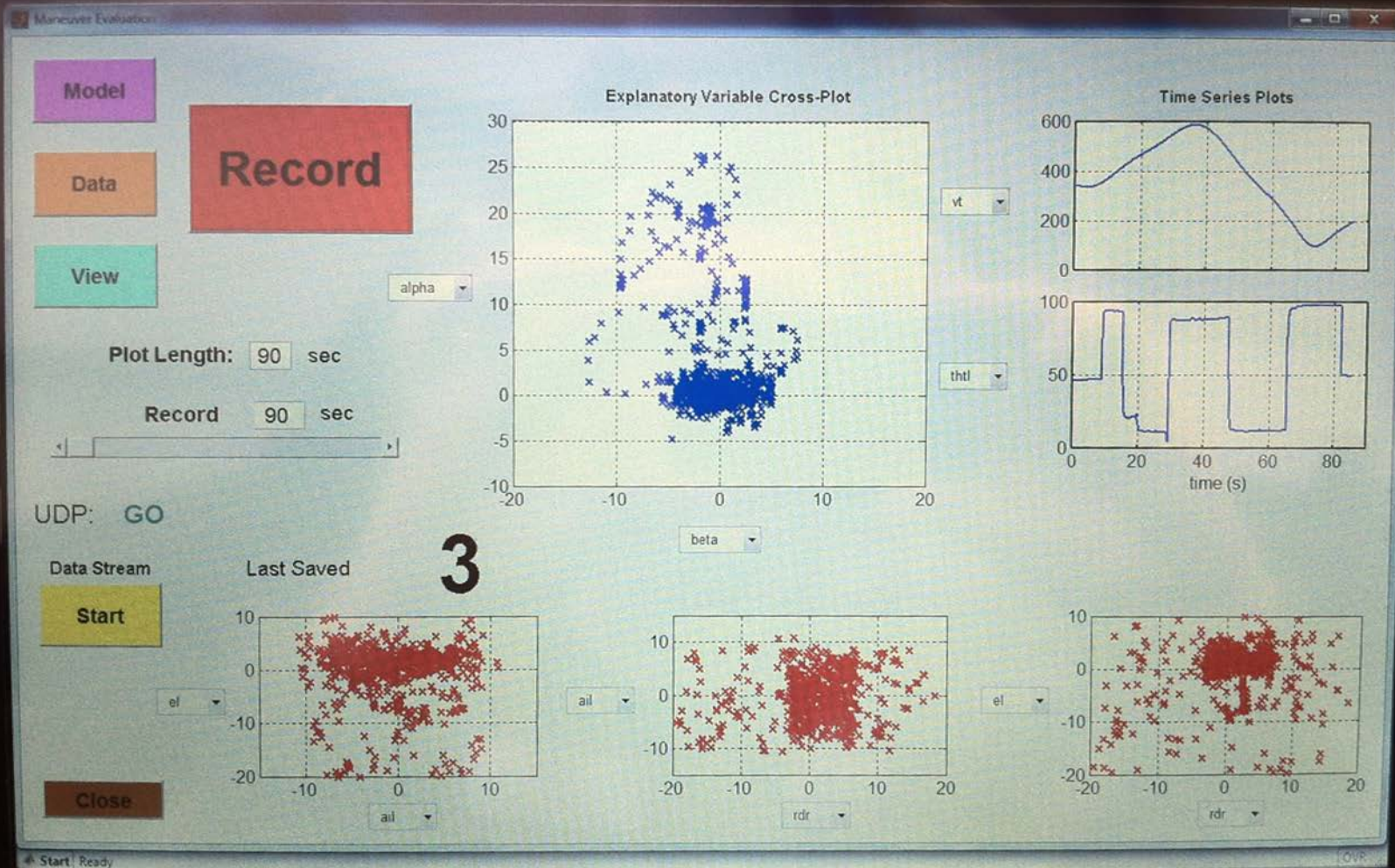
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Flight Test Maneuver Information

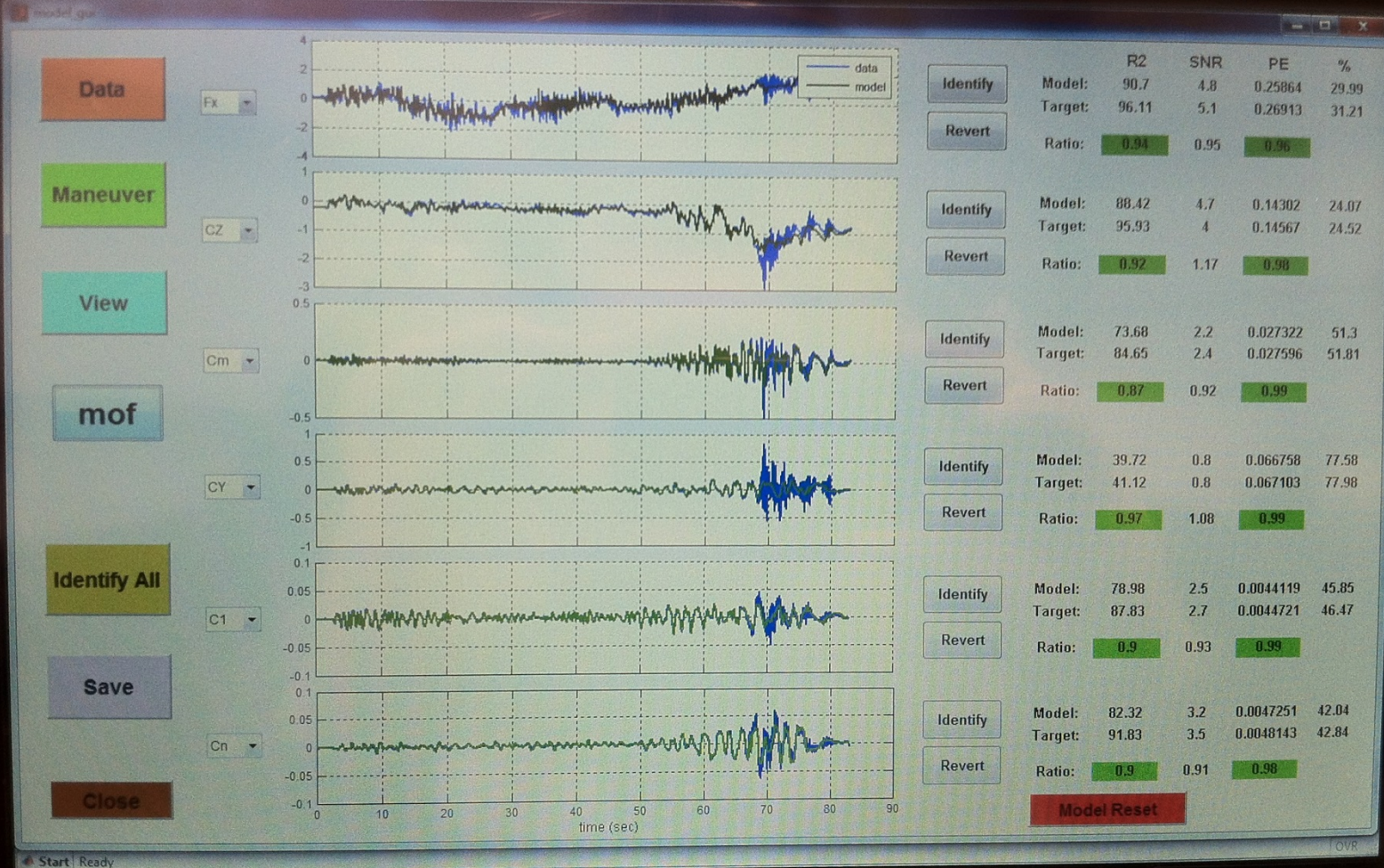
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Modeling Results

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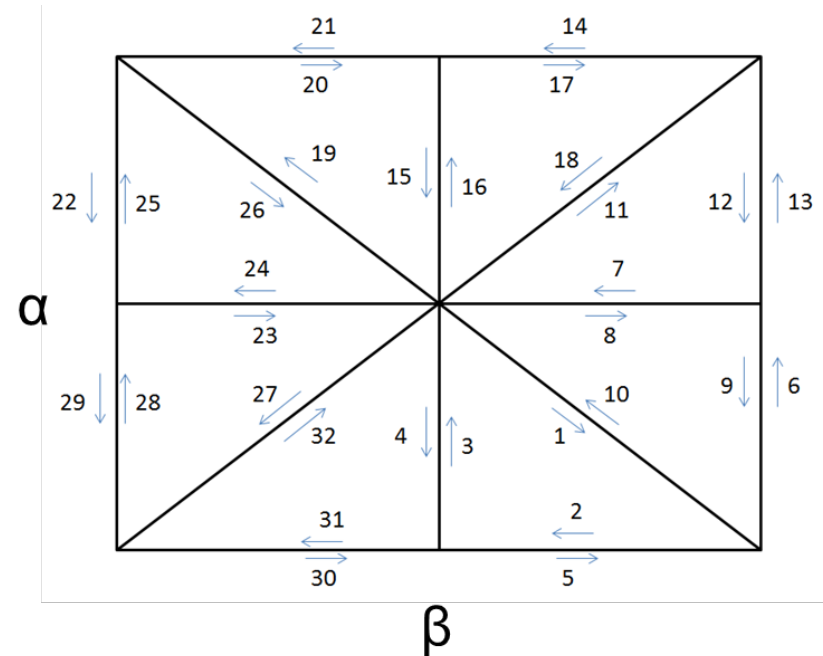
Real-Time Demonstration





L2F Aerodynamics Modeling/L-59 Model Testing

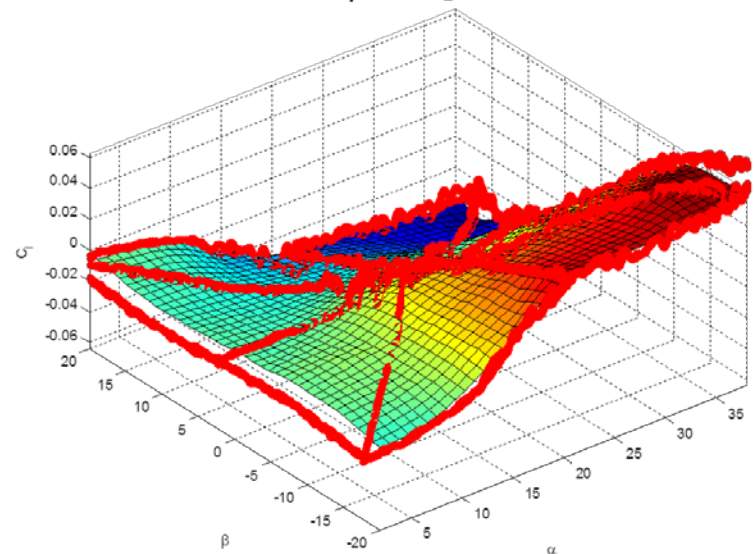
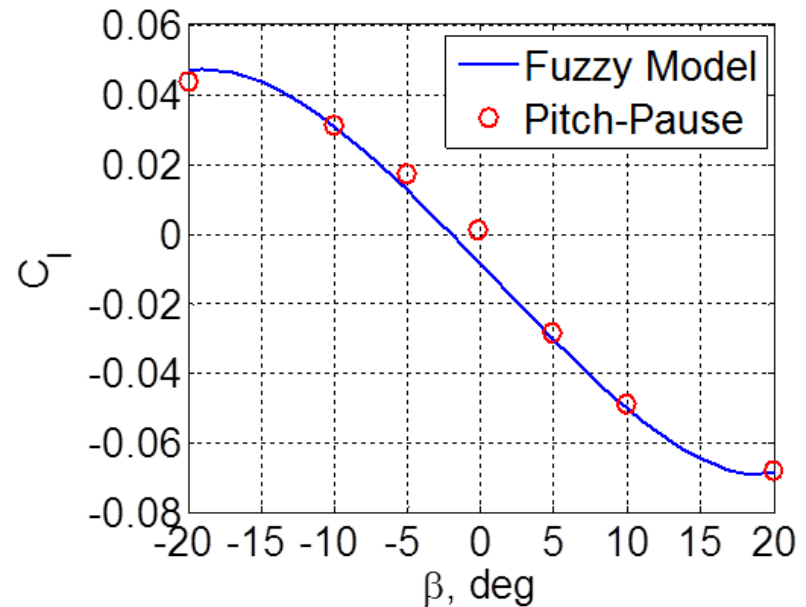
- Objective: Rapid, high-fidelity, nonlinear aerodynamics model generation
 - Large envelope
 - Control effectiveness
- Proof of concept using face-centered “crossbox” test pattern
- R/C L-59 model tested in 12-Foot Low-Speed Tunnel
 - Development of technology and modeling techniques
 - Cheap/expedient model





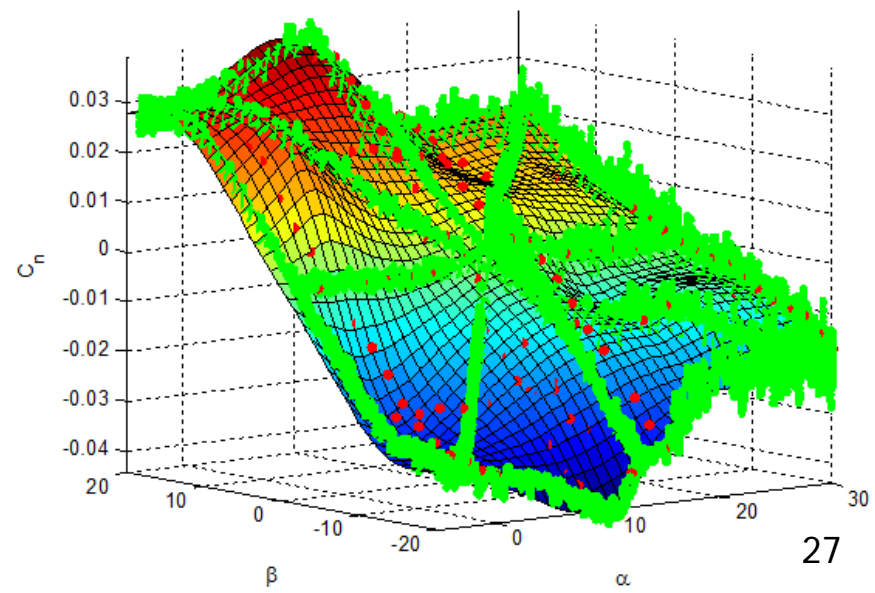
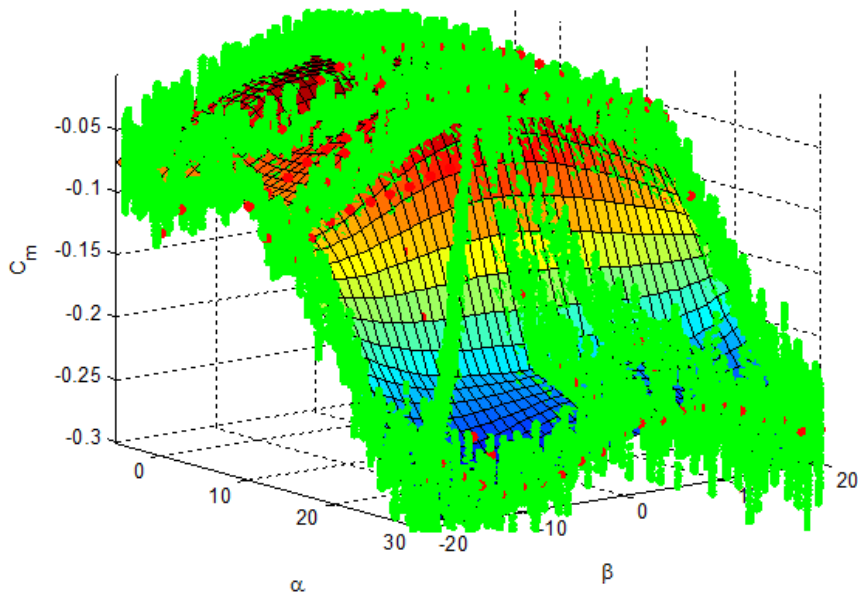
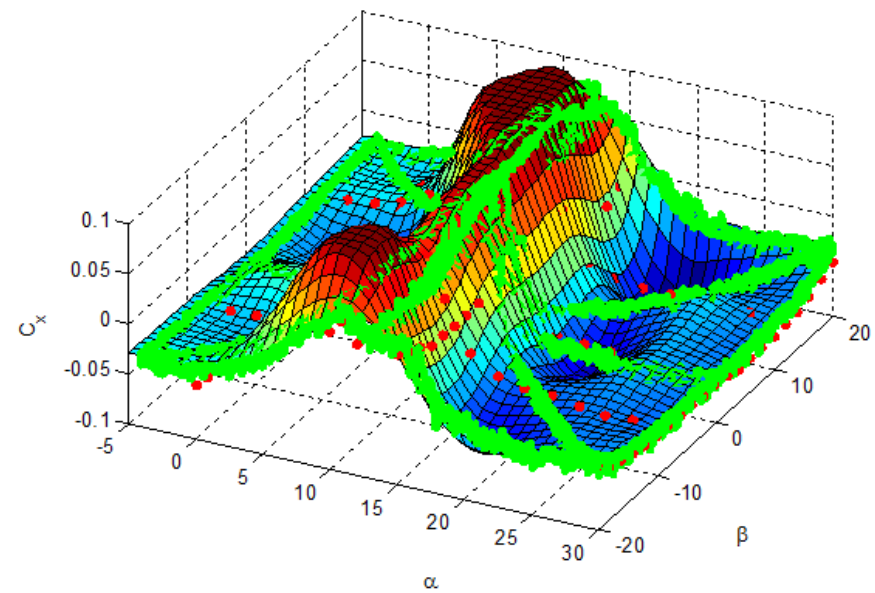
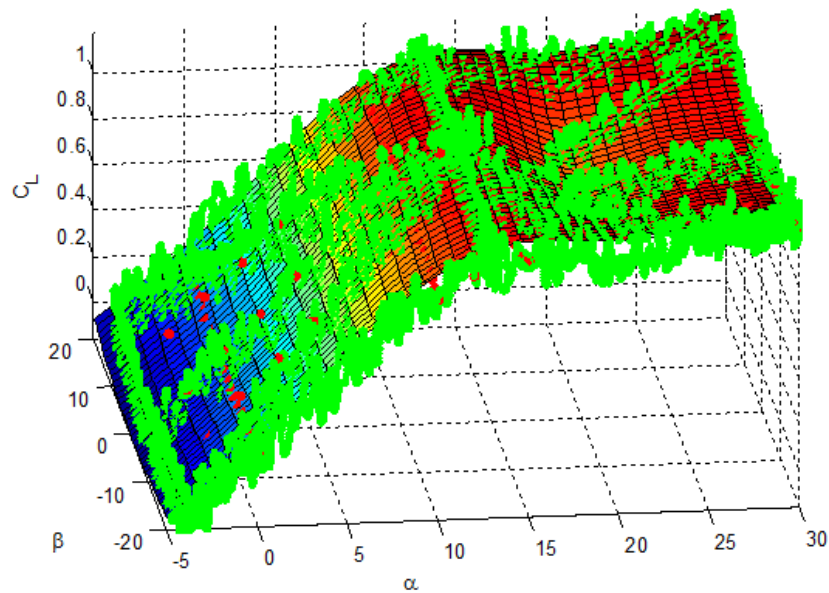
L2F Wind Tunnel Test

- Continuous Motions
 - α and β
 - Control deflections
- Modeling with Fuzzy Logic and Orthogonal Polynomials
- Comparisons with Conventional Data Approach





Other Model Examples





Wind Tunnel Application Plans

■ Plans

- Develop autonomous model assessments
- Develop autonomous decisions on where more data is needed
- Develop/demonstrate autonomous database generation
- Extend to dynamic testing (roll motions) combined with static and controls



Summary

- New testing and modeling approaches shown can achieve high-quality results with significant reductions in cost and time:
 - Automated onboard modeling, real-time flight maneuver guidance
 - Improved flight crew awareness during flight tests
 - Reduced number of flights and maneuvers required
 - Rapid simulation development or updates for modified aircraft
 - Efficient, automated high-fidelity nonlinear aerodynamic modeling throughout normal and extended flight envelopes – flight or ground testing applications
- A step toward developing self-learning and adaptive aerospace vehicles – airplanes, helicopters, UAVs, spacecraft
 - High fidelity nonlinear models of current aerodynamic characteristics provide a tool for effective control design or adaptations in flight
 - Tool for identifying changes for safety or performance improvements
- More work still to be done to refine and validate the approach, but key aspects can be incorporated into flight test projects now



Questions?





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