

Lessons from Modeling Flexible Aircraft for Active Flutter Suppression

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Active Flutter Suppression (AFS)

Reducing structural weight

- Standards require flutter to be 115% of maximum airspeed (V_{ne})
- Control system could be used to increase flutter speed without structural weight

Addressing design issues

- Predicted flutter speed can drop as design matures
- AFS could avoid expensive redesigns

X-56A Multi-Utility Technology Testbed

- Designed for testing active flutter suppression
- Flexible wings have unstable flutter modes
- AFRL Funded
- Lockheed Martin Built
- NASA has flown slightly past the flutter speed





Objective

Generate models useful for the design and evaluation of control laws for active structural control and flutter suppression that are able to accurately predict body freedom flutter.

For design

- Form of the models
 - State-space models
- Interpolation between flight conditions
 - Full envelope design

For evaluation

- Uncertainty
- Piloted simulation

Prediction

- Physically based models
 - Using information typically available before flight
- Predictive accuracy insufficient/inconsistent
 - Using flight test results to determine where we went wrong



Challenges with Aeroservoelastic Modeling

Challenging Dynamics	<ul style="list-style-type: none">• Unstable• Closed Loop
Model Complexity	<ul style="list-style-type: none">• High Order• Multiple-input• Multiple-output
Nonlinearities	<ul style="list-style-type: none">• Difficulty in prediction• Difficulty in interpreting
Uncertainty and Noise	<ul style="list-style-type: none">• Turbulence• Limited Knowledge• Variability



Types of models being used

NDoF model

- Preflight models
 - Did some updates from stiff wing data
- Very high order
 - ~300 states
- Integrated into piloted simulation

Low Order Equivalent System (LOES) models

- Transfer functions
 - Different from handling qualities LOES
- Lower order system
 - Do have structural modes
- Flight derived
 - Time domain system ID

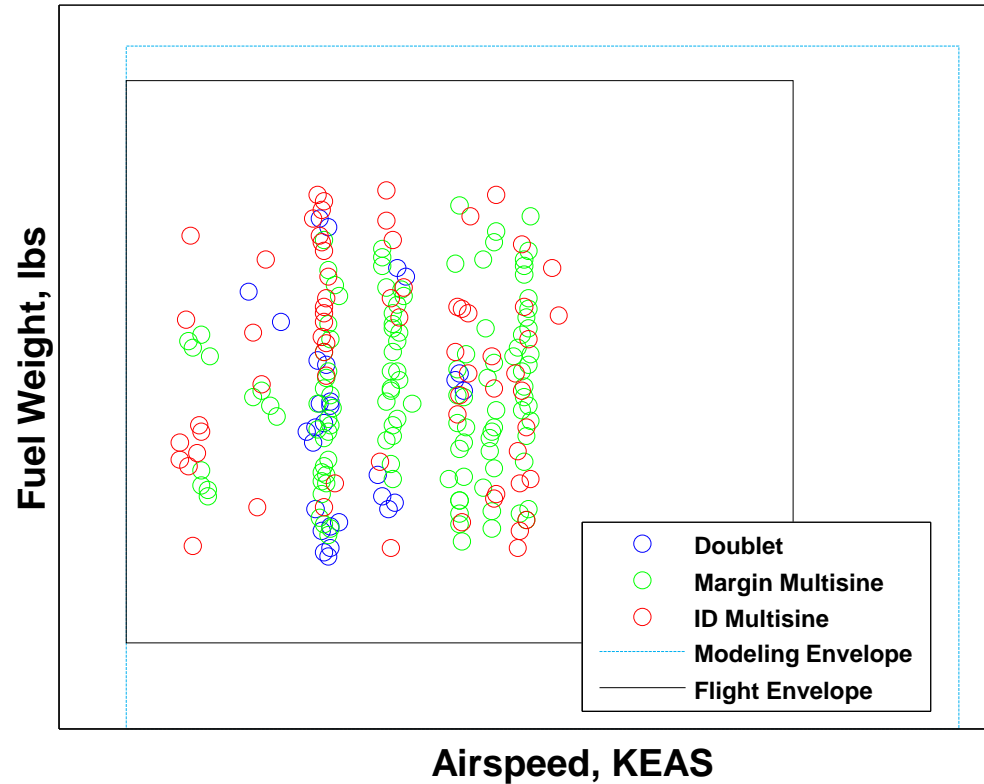
Parameter estimation models

- Lower order
- Flight derived
- More physical parameters
- Still in development

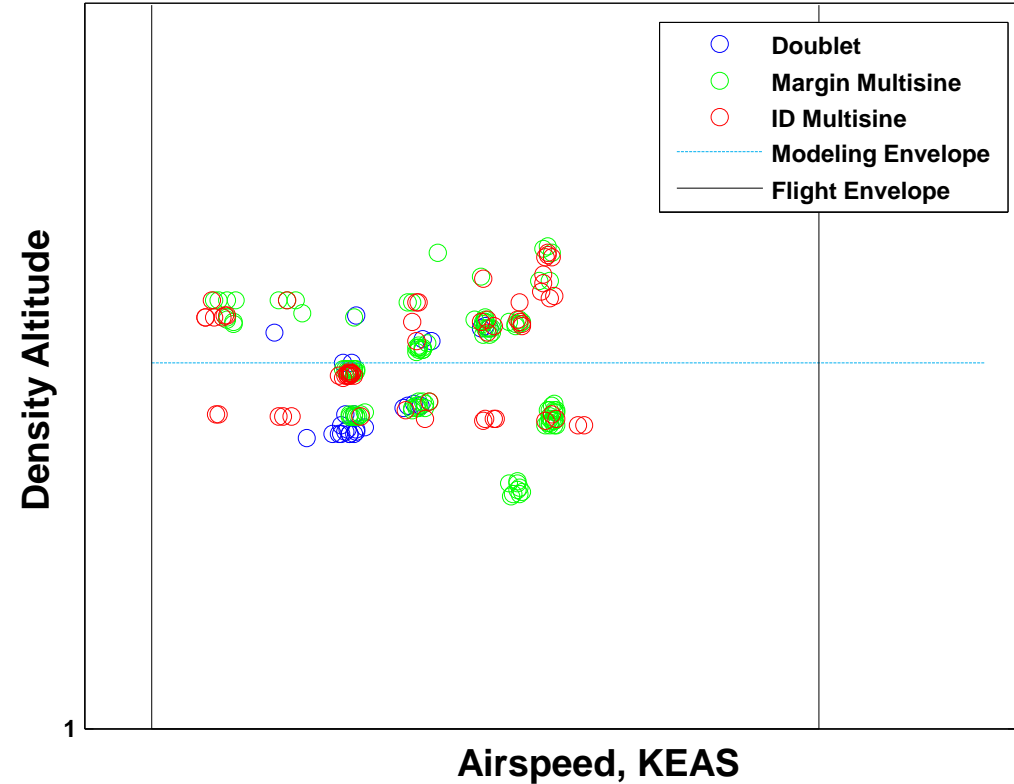
Frequency responses

- Flight derived
- Non parametric (high order)

Maneuvers Used



- Use air density in place of altitude
 - Small effect for low subsonic conditions
- Use ID multisines for updating
 - Better excitation for system ID



- Use margin multisines for validation
 - Keeps data independent of tuning
 - Plenty of test points for comparison
 - Compares specific input/outputs of interest

Model Complexity: “Curse of dimensionality”

HIGH ORDER MODELS

MIMO SYSTEM



Aeroelastic models have lots of parameters

- Parameters arranged in a matrix

$$\begin{bmatrix} A & B \\ C & D \end{bmatrix}$$

- 22 Outputs
 - 12 Inputs
- Estimate of parameters for multiple model types

	Rigid longitudinal model	Parameter ID models	Aeroelastic Plant	Full Aeroelastic Model
States	4	6	155	234
Max parameters	416	504	33,807	62,976
Typical parameters	135	270	10,299	14,958

- System order causes many parameters
 - Inputs and outputs also mean a lot of parameters
 - Physical modeling requires more parameters than identifiable from flight

Preflight Flutter Model Tuning

Flutter model uses a lifting surface method

- Similar to vortex lattice
 - potential flow without boundary layer or thickness effects
- Good for unsteady aerodynamics
- Poor for steady coefficients

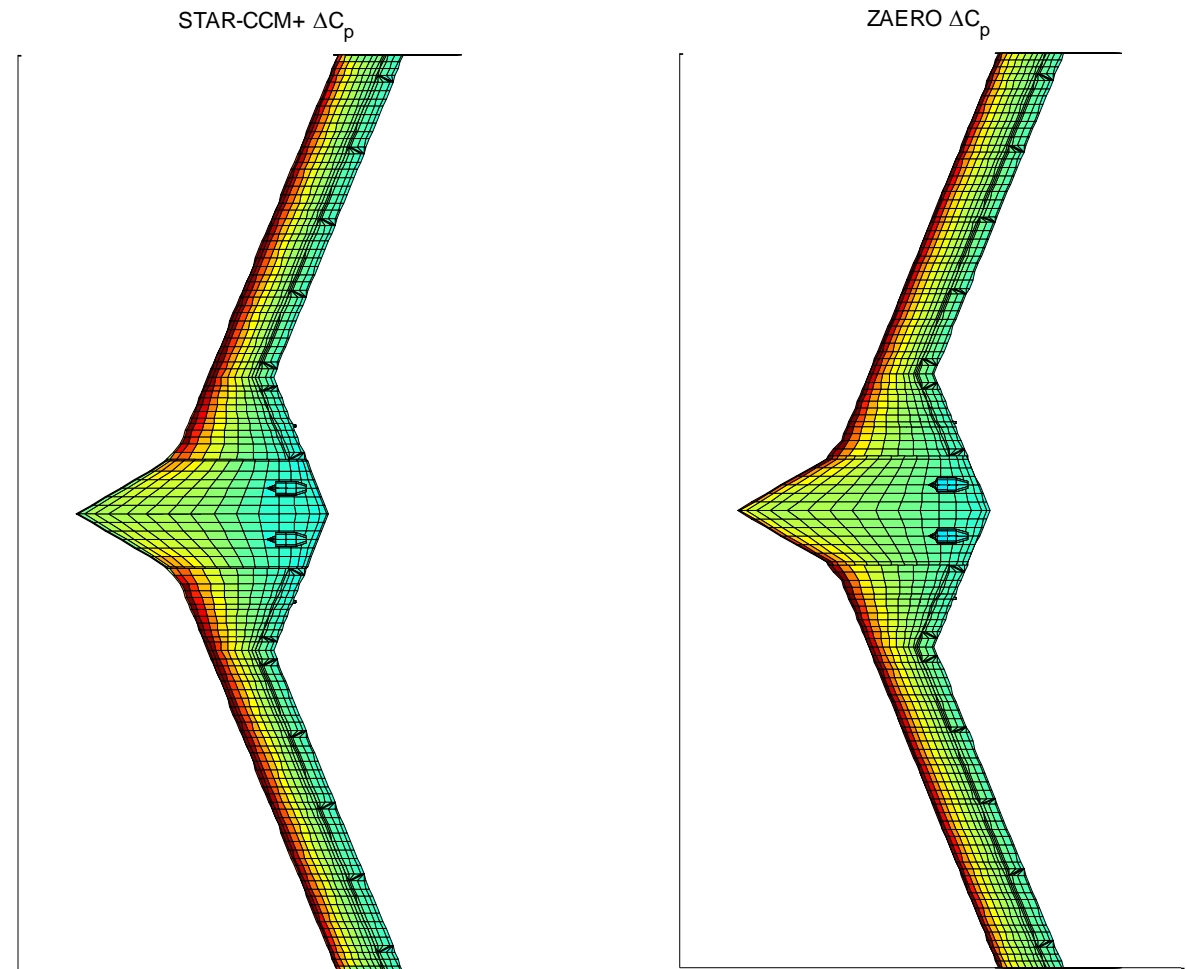
Refining flutter aerodynamic model

- AIC matrices relate local velocity to pressures

Common techniques exist for refining these matrices

- Tuning to match wind tunnel/CFD results
- Effectively changing the shape to reflect the boundary later and thickness.
 - Downwash correction
- Fairly easy to implement
- Fairly easy to create problems

Currently only matching steady coefficients



Issues in flutter model tuning

Correction reflects error in relative change in local velocity

CFD/Wind tunnel includes physics not in potential flow models

- Cannot tune to match physics not in the model
- Requires replacement of coefficients

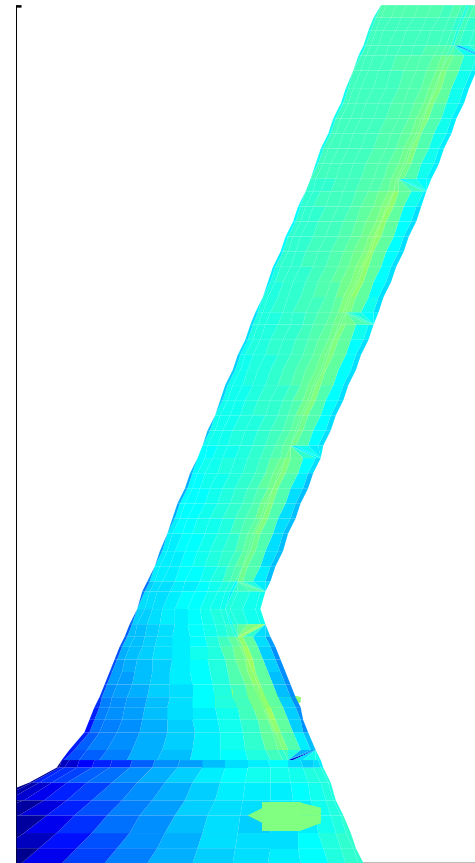
Coefficients may not be consistent

- Matching lift and moment may require unrealistic center of pressure
 - Mostly an issue in control surfaces
- Causes unrealistically large corrections
- Can only match a limited number of coefficients

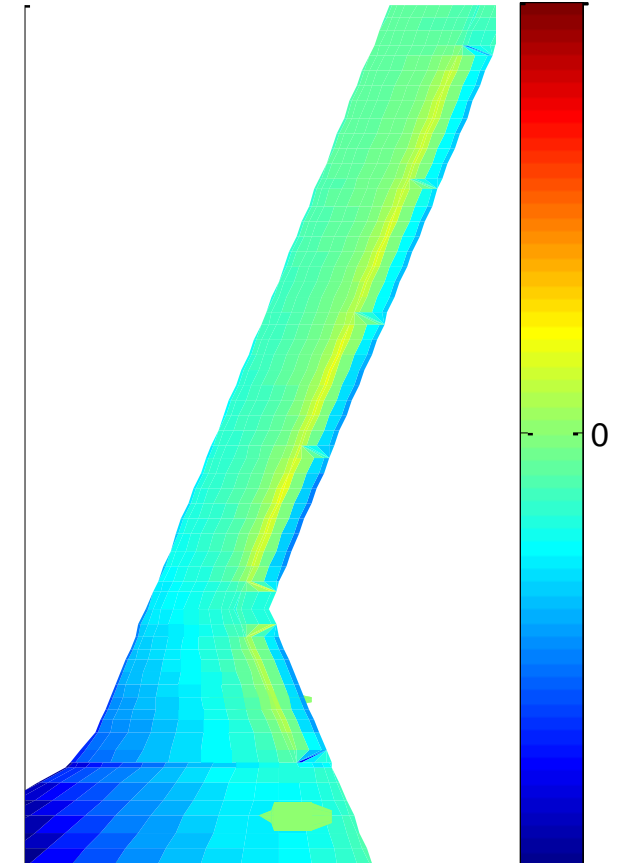
Smoothness of corrections

- Giesing, Kalman, and Rodden used weighted RMS
 - Limits total variation
 - Large changes between neighbors is possible
- We have implemented smoothness based on difference between neighboring panels

CFD Based Correction Factors



Final Correction Factors



Direct Tuning with Flight Data

Low order PID results have different model structure

- Lower order means the parameters are biased
- Requires an adjustment to parameters

Directly adjusting parameters used in model generation

- No adjustment needed for implementation
- Full envelope correction
- Also applied to generate LOES models

Frequency Domain

- No estimation of states required
- No inclusion of control system needed

Output Error

- Using only accelerometers and strain gauges
- Data directly from flight test instrumentation has more consistent timing





Simplifying the Outputs

Outputs are highly correlated

- Each strain gauge contains essentially the same information
- Turbulence will cause signal errors to be correlated

Principle component analysis (PCA/POD/KLT)

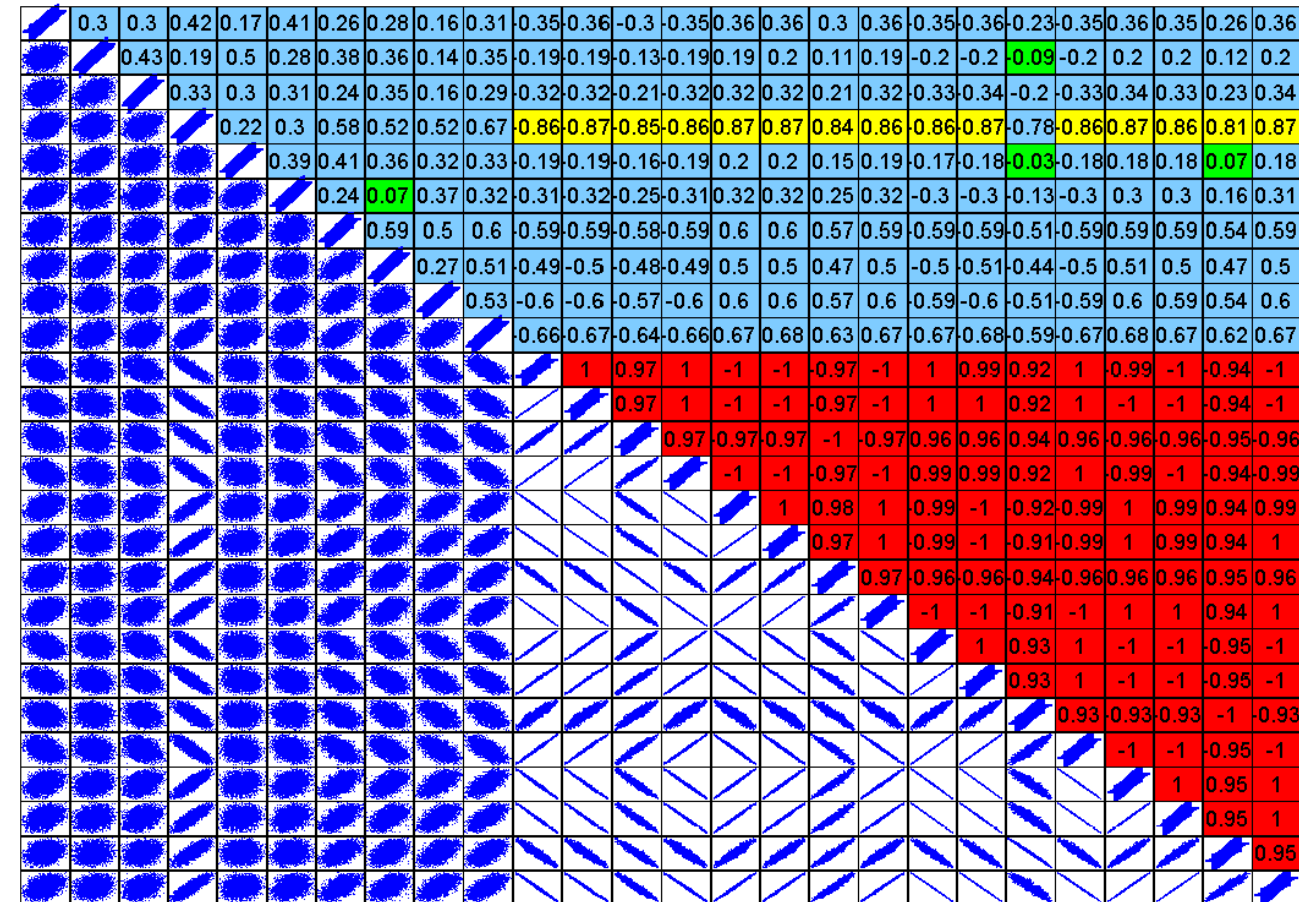
- Combine base on linear relationships
 - Reducing dimension of the outputs
 - Combining outputs to average out the noise
- Similar to generating modes

Reduced rank regression

- PCA applied to outputs

Tested on generating low order models

- Reduces number of parameters fit
- Improves speed (~100x) and accuracy of the fitting



Accels	Strain Gauges
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Direct Tuning with Flight Data

Tuning is improving the fit

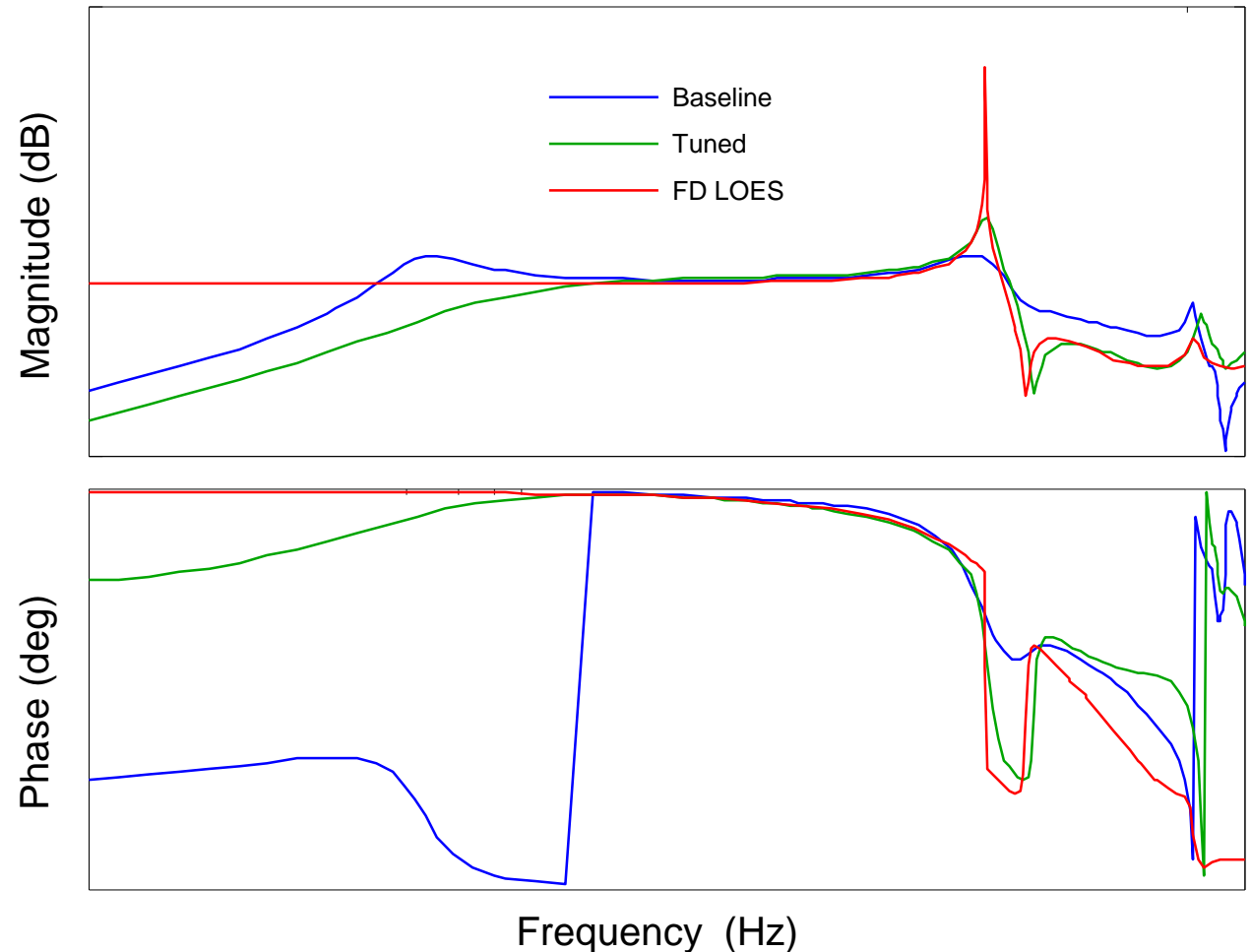
Additional improvement is still expected

- Flutter speed is still high
- Unable to get full envelope correction
- Large corrections

Additional parameters may be needed

- Updating 58 parameters
- LOES models have ~200 parameters
- Inertial Parameters
- Structural Parameters
 - Update material properties, rather than frequencies
 - Mode shapes are fixed
 - Output equations are fixed

Bode Diagram: Wing Flap 4 to Center Forward Accel



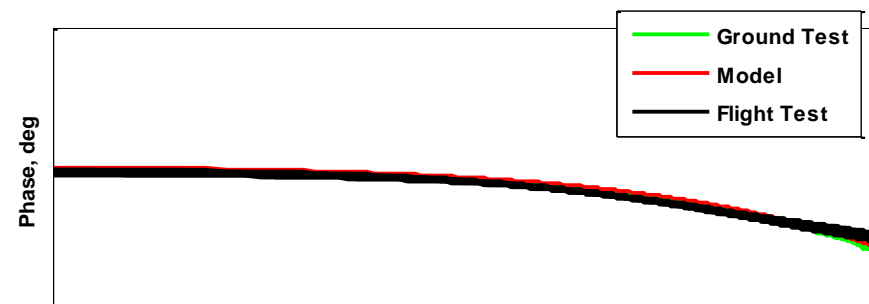
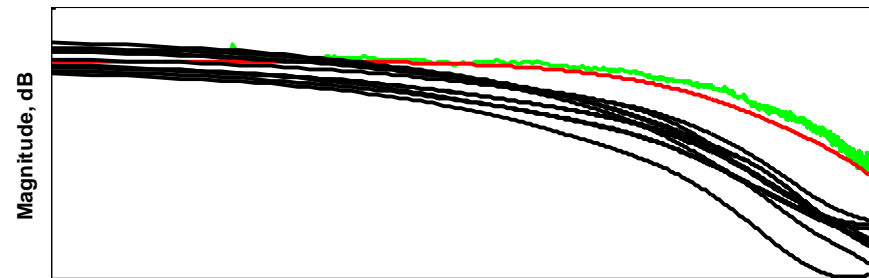
Nonlinearities

ACTUATORS

TAKE-OFF AND LANDING

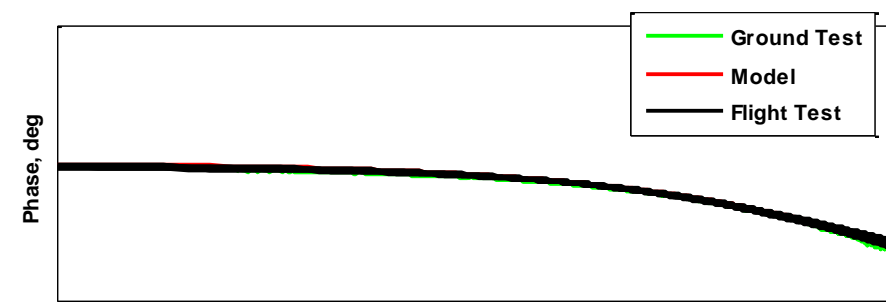
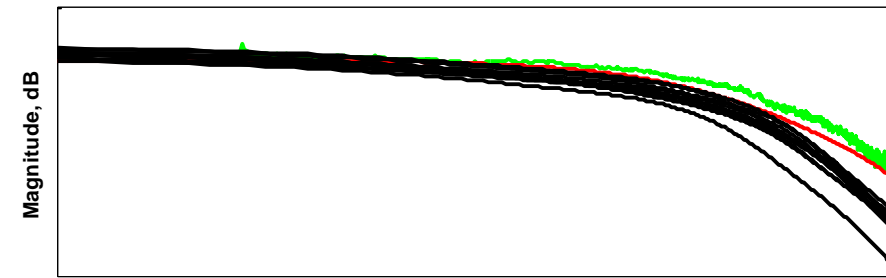
Effect of Actuator Nonlinearities

Medium Magnitude



Frequency, Hz

Large Magnitude



Frequency, Hz

- Actuators do not respond to small commands
- Same flight and input (system ID), different magnitudes (x2)
- Same low airspeed, similar fuel weight
- Smaller magnitude shows more variability, and large error relative to ground test

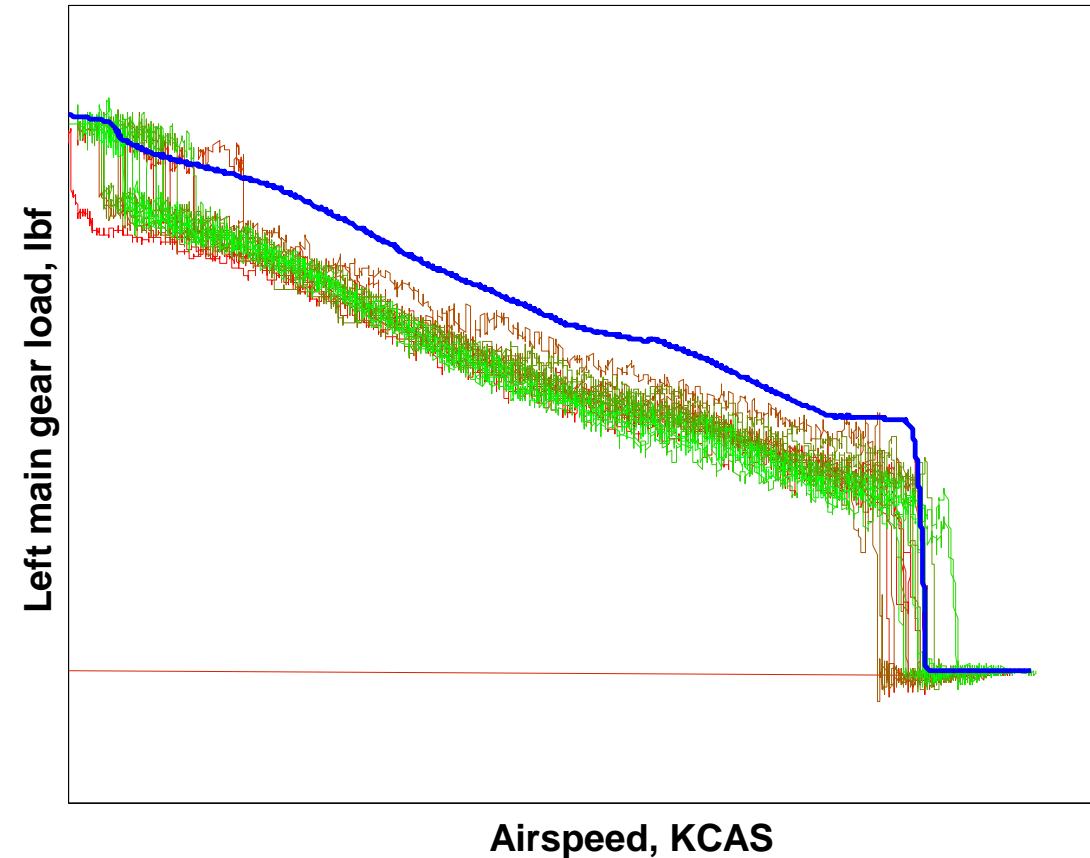
Take-off Simulation

Rapid decrease in main gear loads with increased throttle

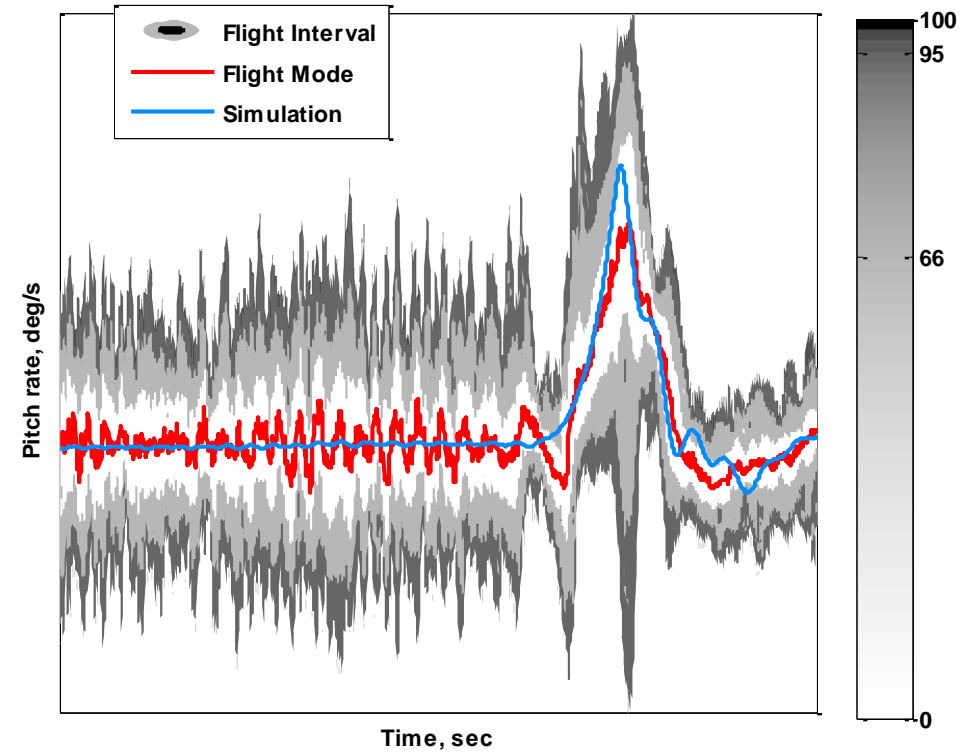
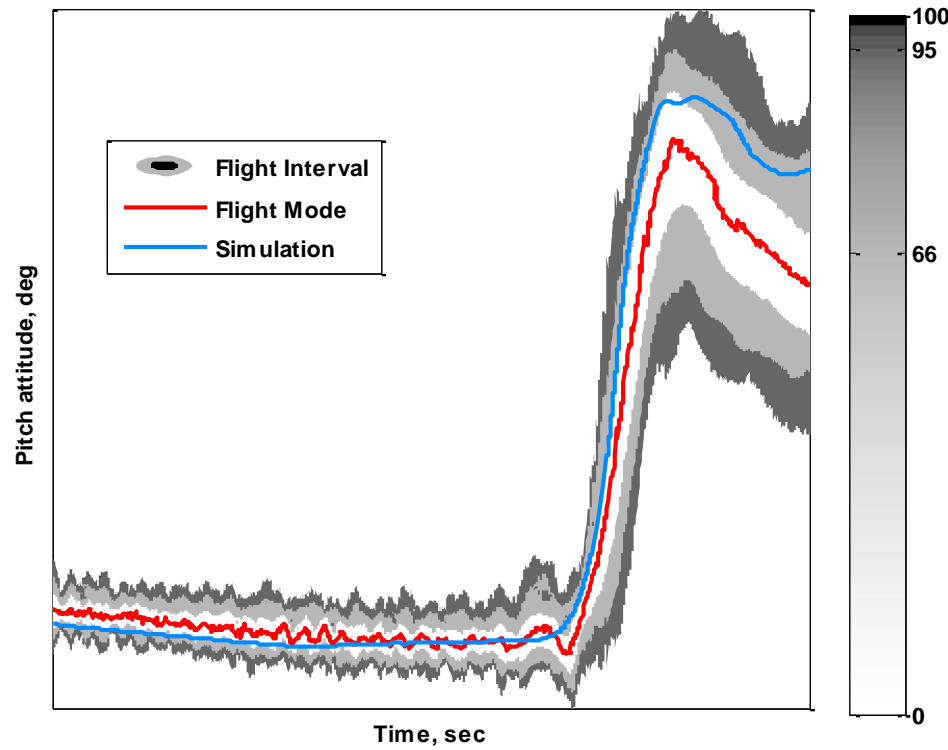
- Not captured in the simulation
- Change in load is ~40% of change in thrust

Simulation predicted higher rotation speed

- Higher rotation speed irritated the pilots
- Simulation used in training adjusted ground effect
 - Not physical
 - Small detriment to accuracy of landing dynamics



Take-off Simulation



Higher rotation speed causes larger pitch up

Flexibility degrades landing damping



- On touchdown significant energy goes into the wings
 - Wings are very poorly damped
 - Reduces effectiveness of landing gear
- Rigid body simulation did not reliably predict the response
- Out piloted simulation included structural dynamics

Uncertainty and Noise

TURBULENCE

LIMITED KNOWLEDGE (EPISTEMIC)

VARIABILITY (ALEATORIC)



Turbulence

First flex wing flight (flight 9)

- Encountered light to moderate turbulence

Pilot perception

- Had simulated similar levels of turbulence
- Response was more stressful in flight

Structural dynamics

- Added to piloted simulation before flex wing flights
- Effects were not added to nose camera motion

Turbulence model

- Used standard model from loads and handling qualities
 - These primarily excite rigid body motion
- Higher frequency turbulence caused more structural motion



Limited Knowledge (Epistemic Uncertainty)

Uncertainty often viewed as a lack of knowledge

- More testing could improve knowledge and reduce uncertainty
- More testing not always necessary or practical

Unrelated parameter uncertainty

- Large number of parameters
 - Caused unrealistically large uncertainty in output
 - Error in parameters should be related
- Down select parameters based on engineering judgement
 - Missed parameters that are important

Mu-analysis was very appealing

- Computational cost was excessive
 - To many parameters to examine
- Frequency of parameter occurrence
 - Mass parameters and air density effect many parameters
- LFT format was still useful for Monte Carlo

Flight models show what output uncertainty should be

- Models were sufficient for controller design, once the uncertainty was known



Variability (Aleatoric Uncertainty)

Landing Sensitivity Analysis

- Some uncertainty is inherent, and cannot be reduced

Examined sensitivity of the response to parameters

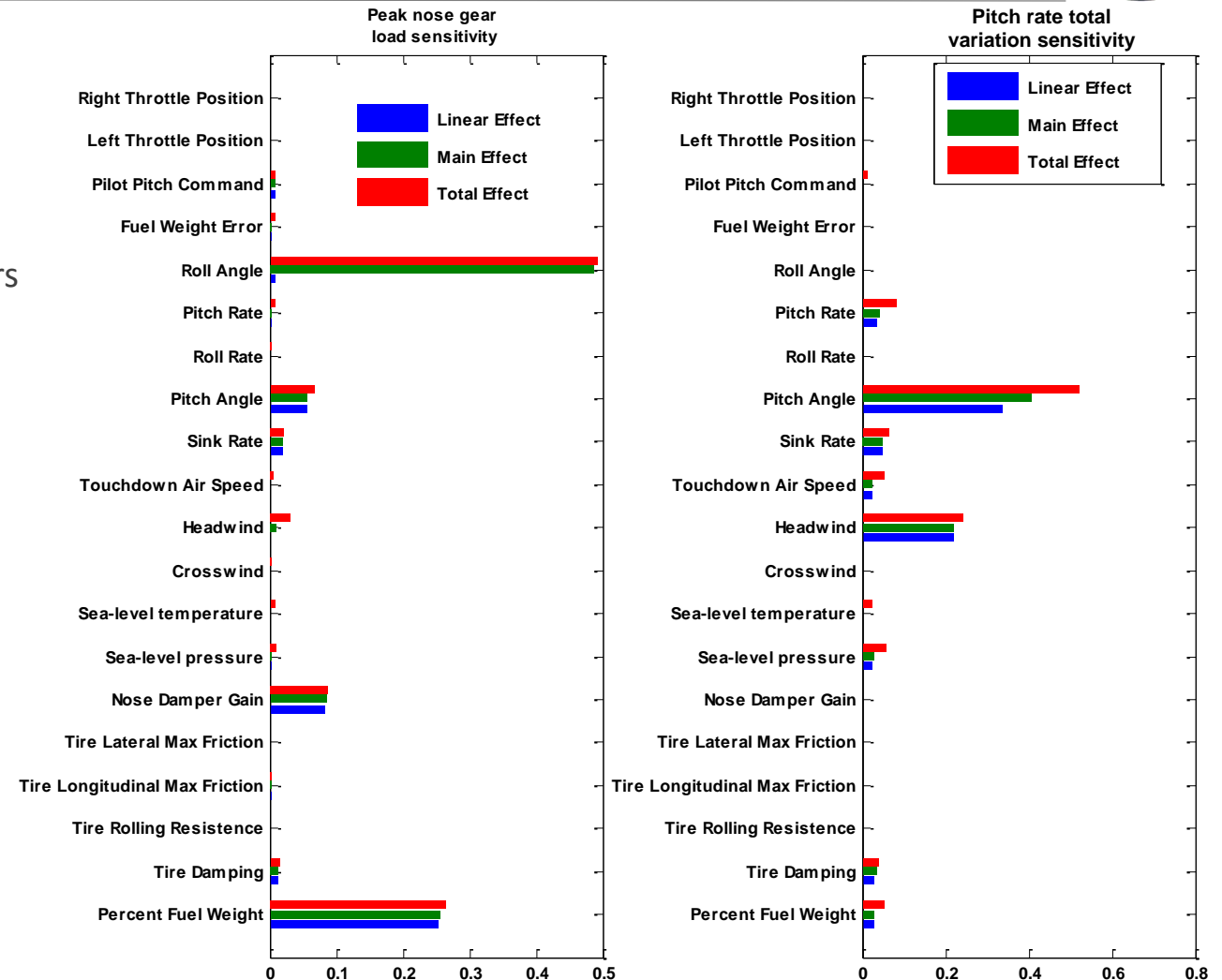
- To many parameter combinations to consider
- Needed to understand interactions between parameters
- For 26 parameters, $2^{26} \approx 10^8$ possibilities
 - >2 years for a 1 second simulation

Monte Carlo/Polynomial chaos expansion

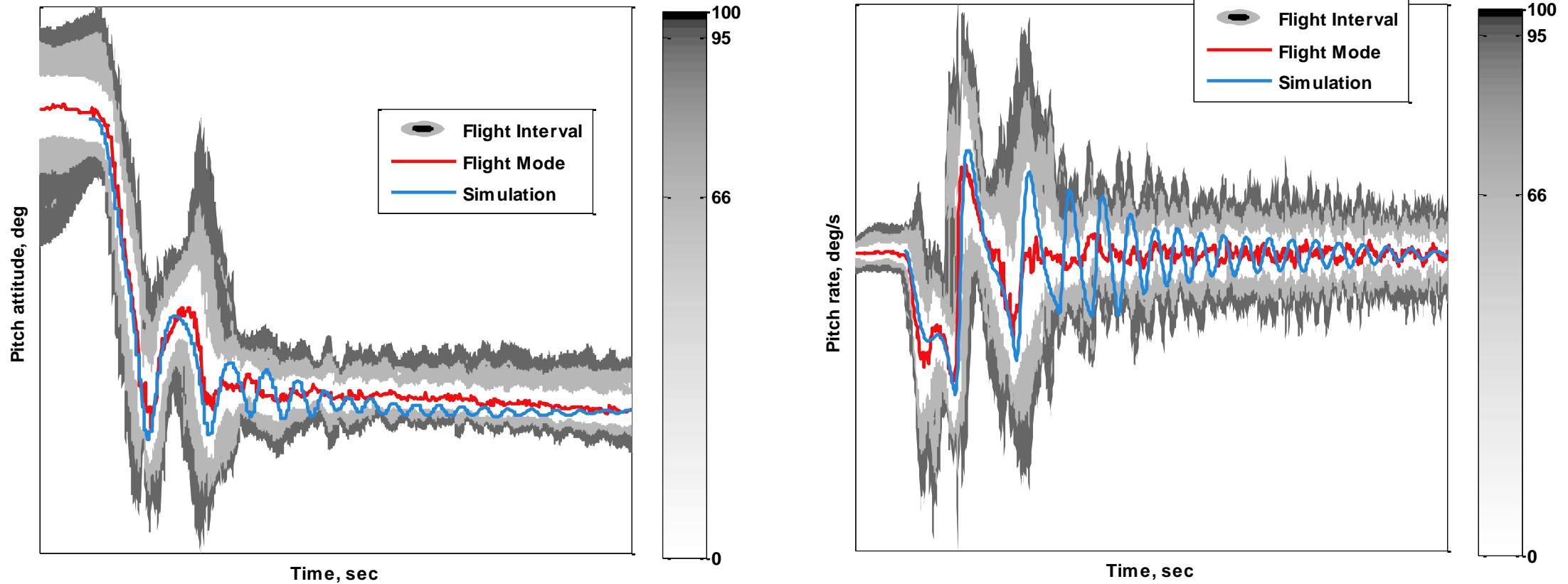
- Generate surrogate model
- Polynomials orthogonal with respect to parameter probability distribution

Used multiple types of sensitivity

- Linear effect: traditional linear sensitivity
- Main effect: direct nonlinear effect
- Total effect: Includes interaction between parameters



Landing Dynamics



Simulation is capturing initial dynamics

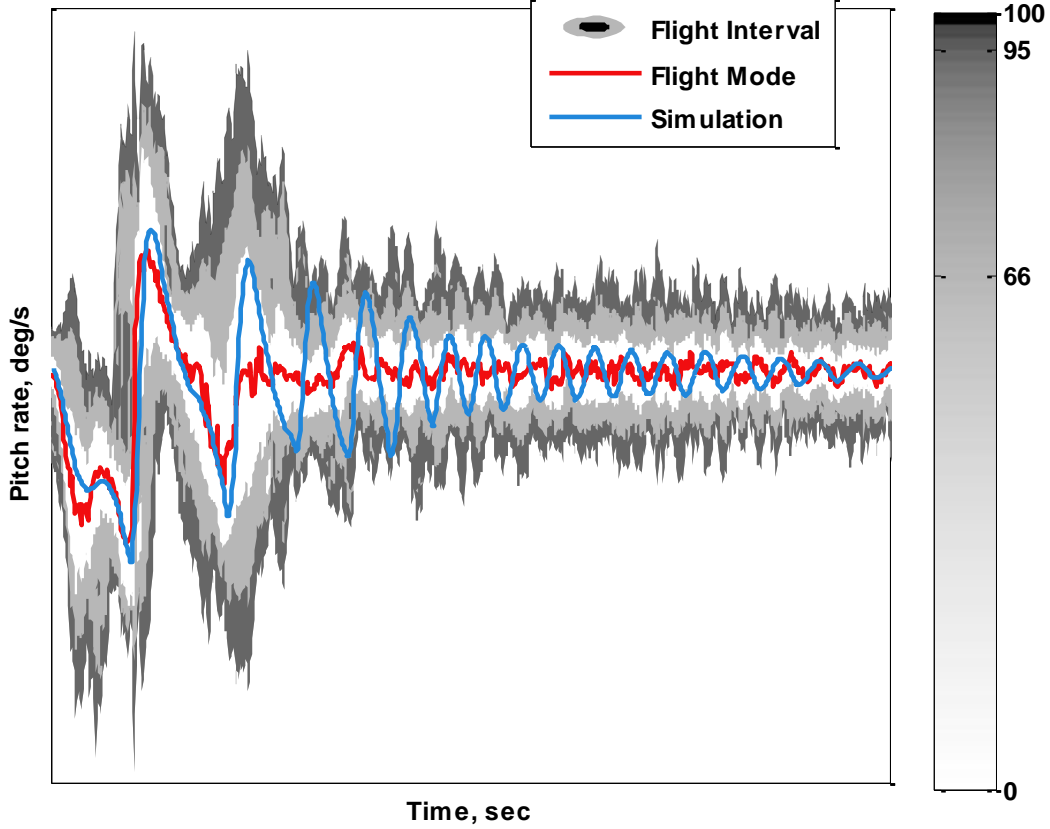
- Nonlinearity of ground contact makes comparison difficult

Having a piloted simulation allowed for the development of an effective landing technique.

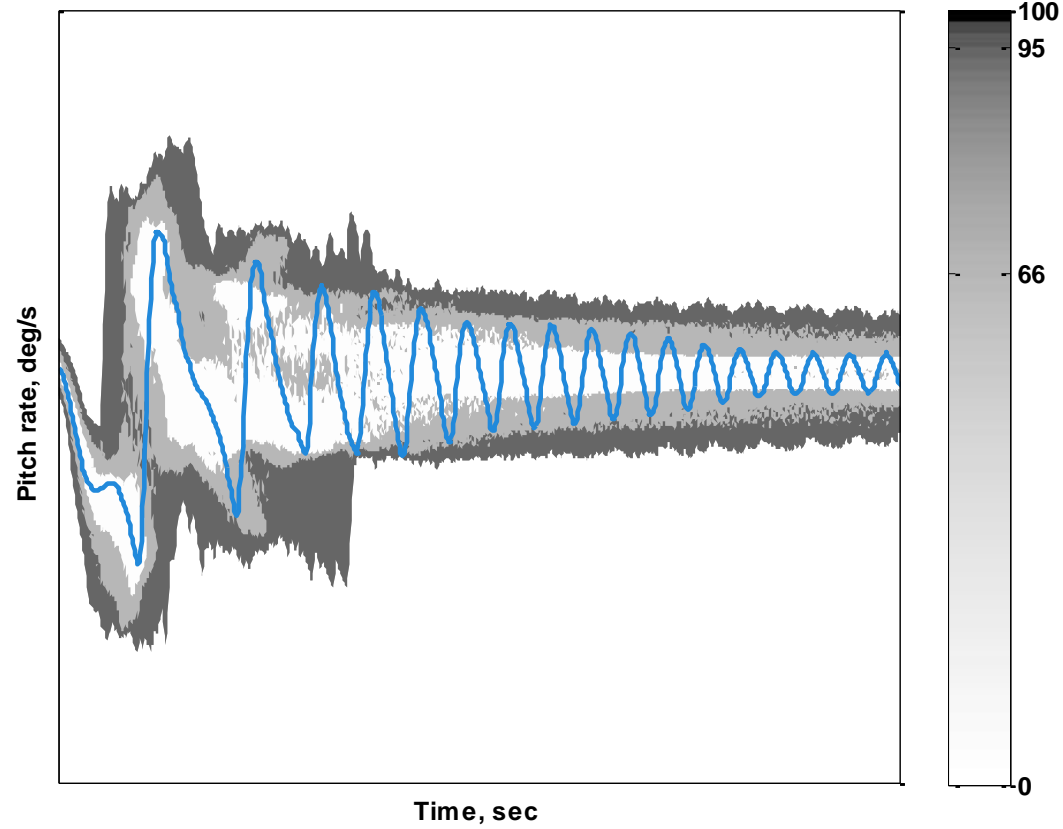


Landing Monte Carlo

Flight



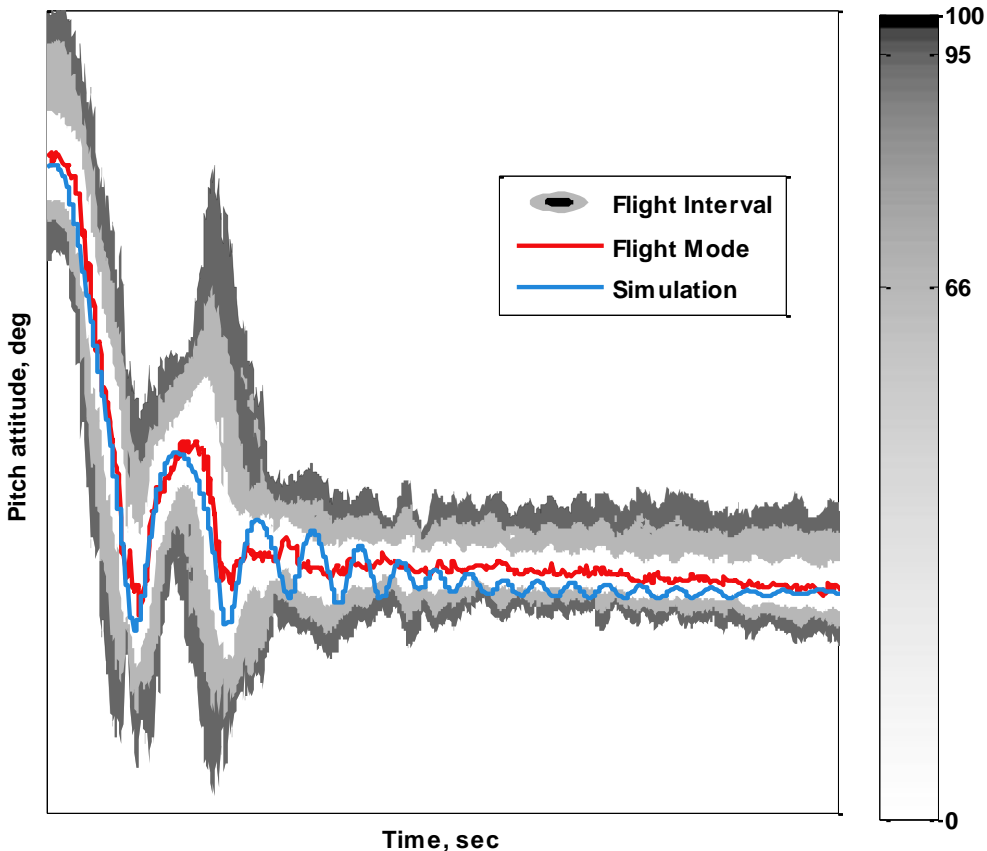
Simulation



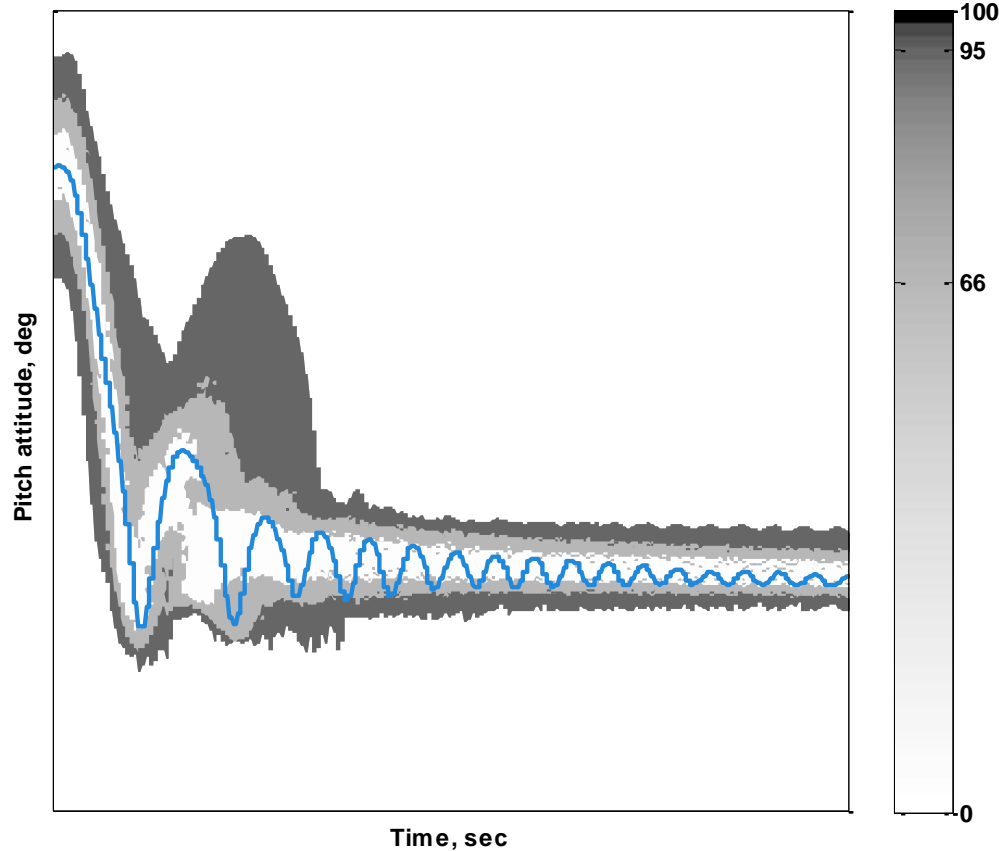


Landing Monte Carlo

Flight



Simulation



Conclusions

Conclusions

Have generated models of reasonable accuracy

- Used to design control system past flutter
- Used to address and validate takeoff and landing issues
- Challenges do still remain

Developed tools and methods for ASE challenges

- Unstable closed loop system
- Complex dynamics
- Nonlinear behavior
- Uncertainty

