

# Failure Domain Bounding with Applications to Dynamic Systems

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# Favorite Quote from NSES 2011

*“When systems fail, people notice.”*

*Dr. J Michael Gilmore*

# Recent NASA History and Context

- **2003 – CAIB Report (Columbia accident):**
  - Engineering solutions should have included a quantifiable range of uncertainty and risk analysis.
- **2005 – RTFTG Final Report (Columbia accident):**
  - Further compounding the modeling challenge is the fact that the models are deterministic, yielding point estimates, without incorporating any measure of uncertainty in the result
- **2005 – NASA CEV RFP**
  - Design and execute a meaningful risk mitigation program that culminates in a risk reduction flight effort and PDR by the end of calendar year 2008
- **2007 – NASA-STD-7009 for Models & Simulations:**
  - The risk assumed by the decision maker is often misestimated due to inadequate assessment of uncertainties
  - Reports to decision makers of M&S results shall include an estimate of their uncertainty and a description of any processes used to obtain this estimate

# Uncertainty Analysis and Robust Design

***Increase confidence and consistency in aerospace vehicle safety predictions by developing improved methods for quantifying and managing uncertainty***

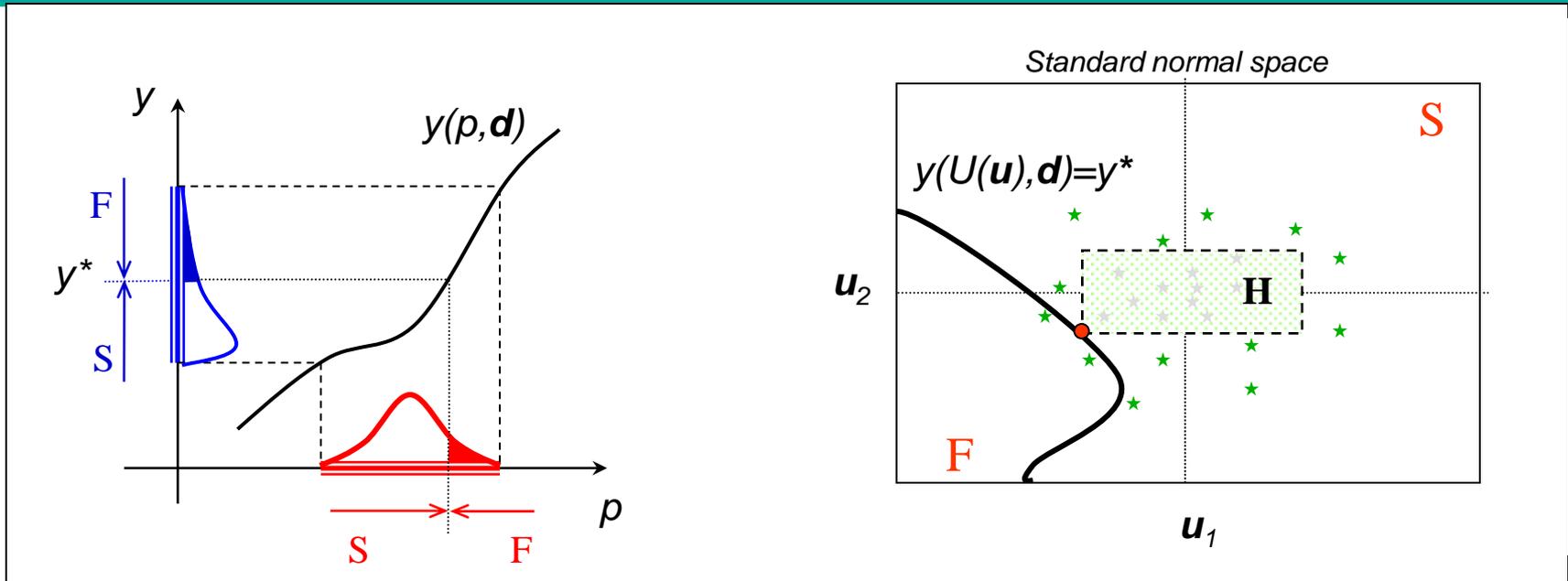
## Quantifying

- Uncertainty Modeling
  - *Model uncertainty based on experimental data, simulations and/or expert opinion*
- Uncertainty Propagation
  - *Given uncertainty models of a system's inputs, how to propagate them through system models, to efficiently evaluate the corresponding system's outputs?*

## Managing

- Robust Design
  - *Generate designs that robustly accommodate uncertainty*
- Uncertainty Decomposition
  - *Identify uncertainties that contribute the most to performance degradation*
  - *Determine the parameters that should (not) be modeled as uncertain*

# Failure Domain Bounding via Homothetic Deformations

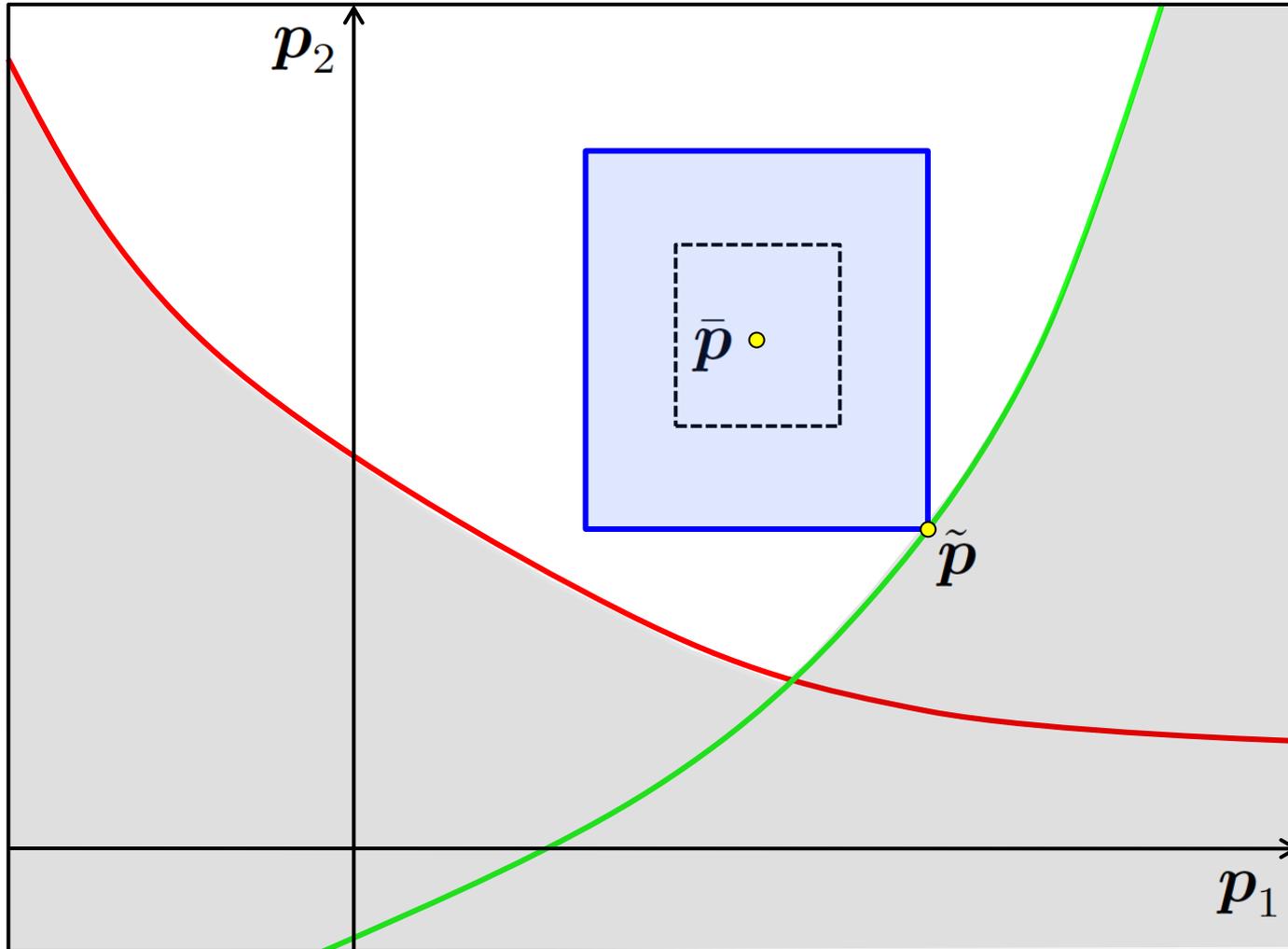


- Monte Carlo sampling
- Homothetic deformations
  - Optimization-based
  - Computationally cheap
  - Analytical expressions for  $P[H]$

## Applications

1. Robustness metric  $\rightarrow$  PSM
2. Upper bounds to  $P[F]$
3. Hybrid Method for  $P[F]$

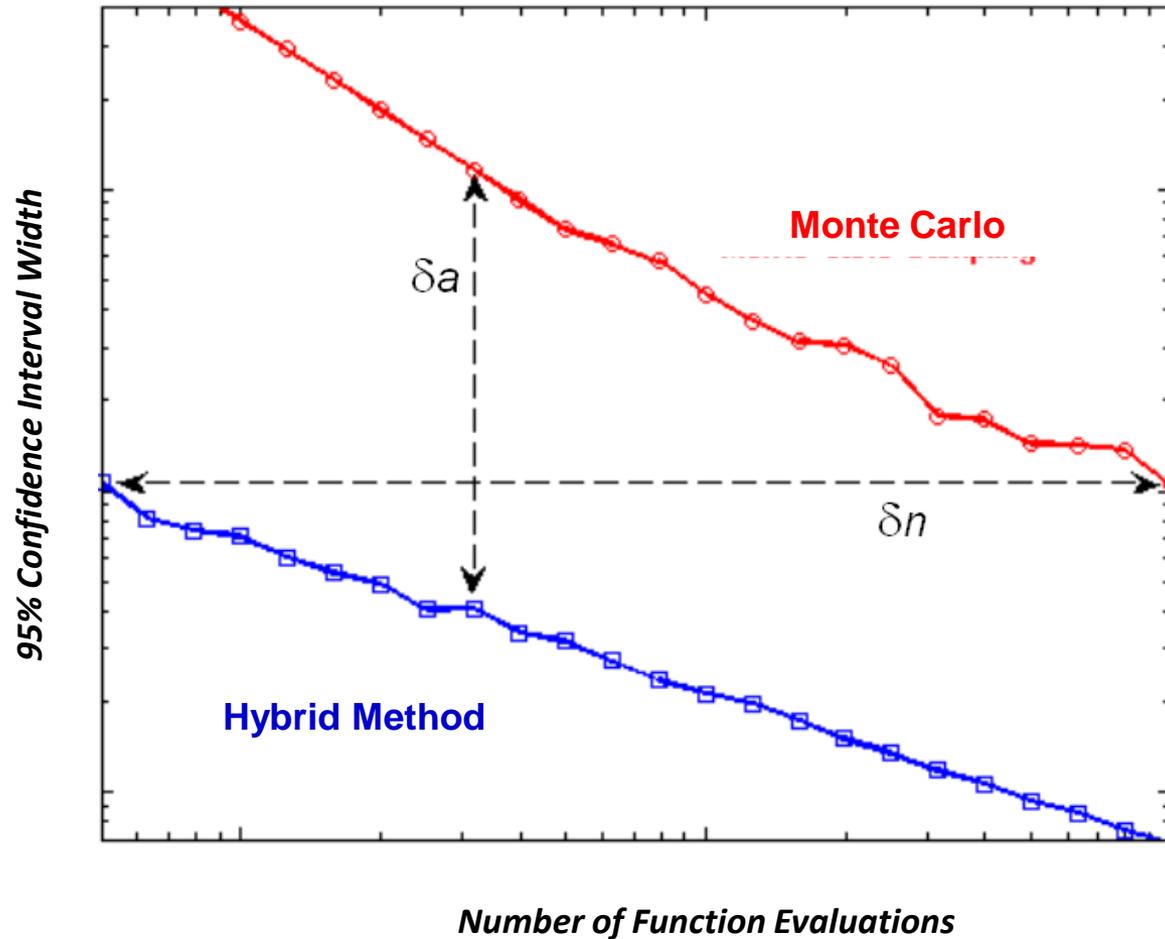
# Analysis: Homothetic Deformations



- *Outcomes: robustness metric, worst-case uncertainty, separation, probability bounds*

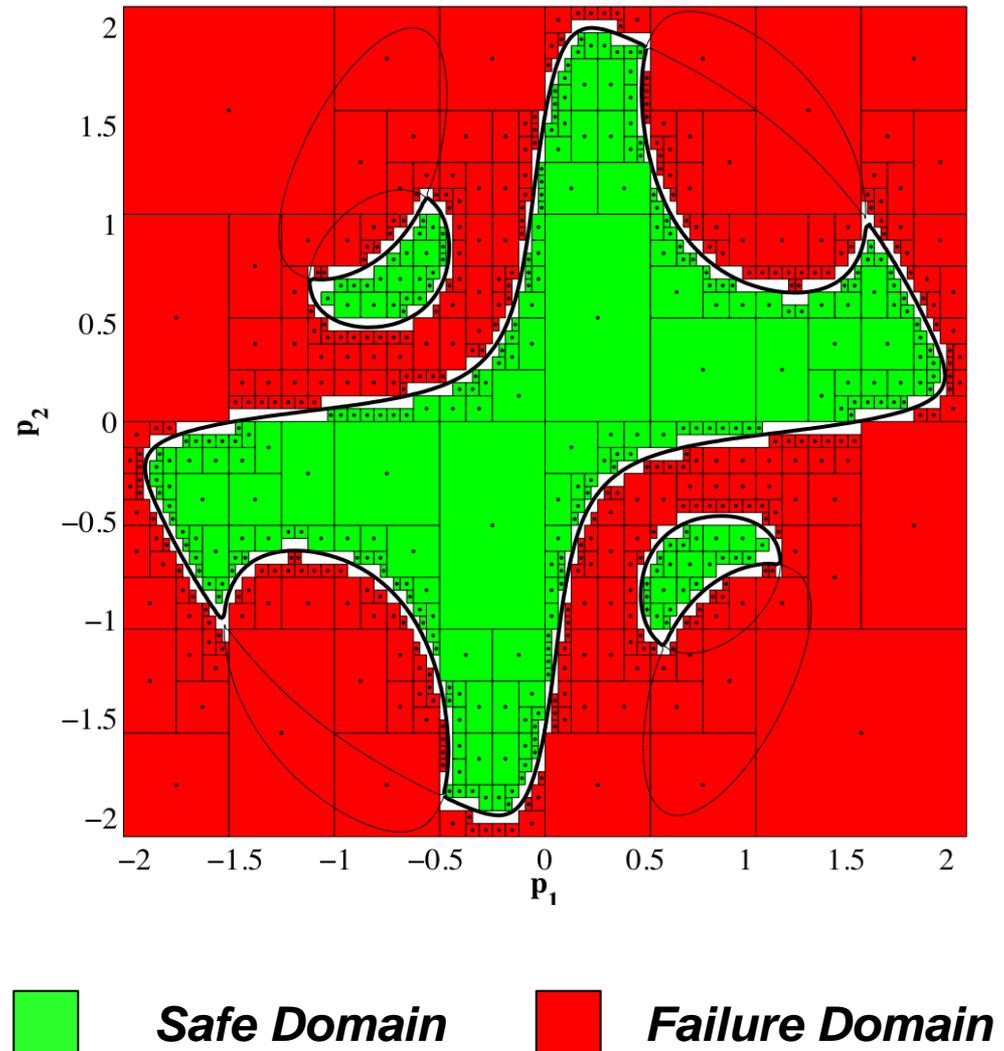
# Efficiency Relative to Monte Carlo

*Hybrid Method vs. Monte Carlo*

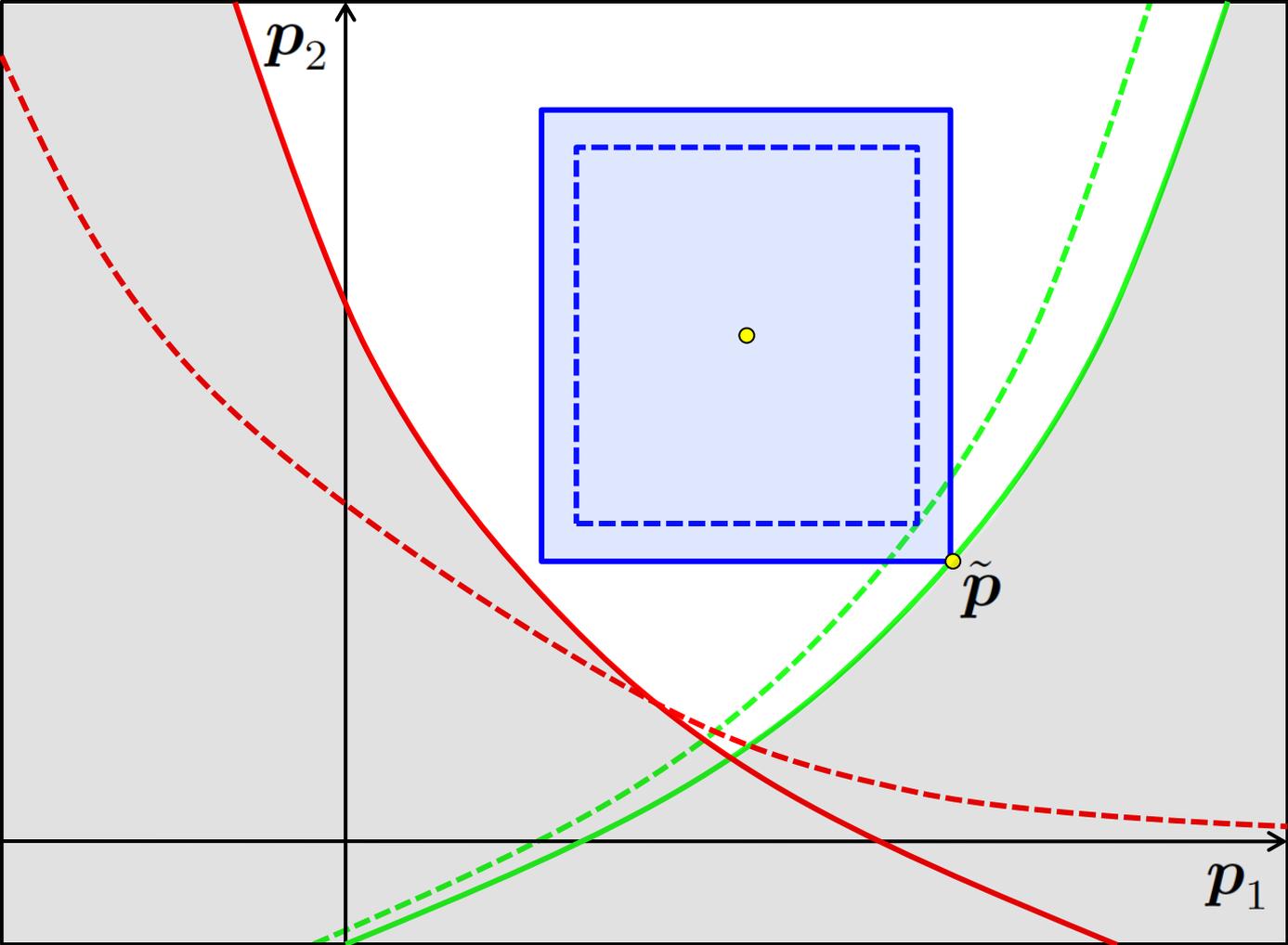


# Failure Domain Approximation

- Builds upon the research in homothetic deformations.
- Yields high fidelity characterizations of complex nonlinear failure domains.
- Utilizes theory of Bernstein polynomials.
- Desensitizes the analysis from assumptions used to model the uncertainty.



# Design: Homothetic Deformations



• *Conflicting objectives, optimally robust designs*

# The Tool Suite: UQTools

**UQTools is a collection of Matlab functions designed to quantify the impact of uncertainty on generic, continuous, parameterized models.**

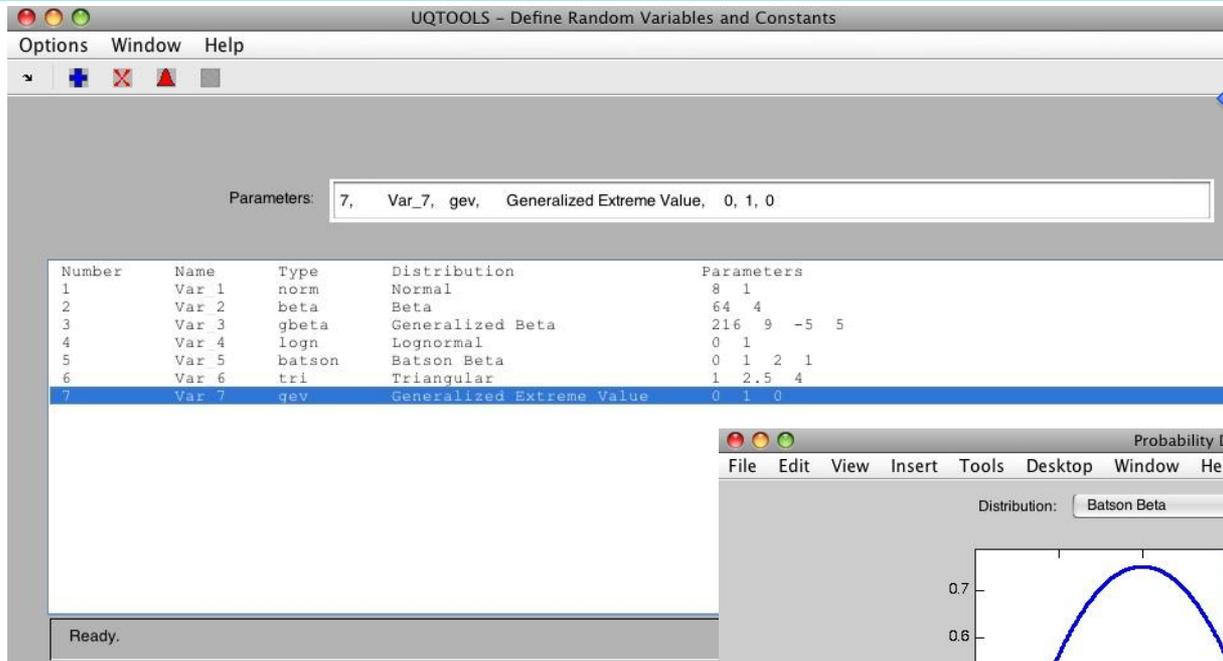
- **Uncertainty Models:**
  - Probability density functions
  - Non-probabilistic sets (hyper-rectangular and hyper-spherical)
- **System Models:**
  - Matlab-callable parameterized input/output maps.

# UQTools Capabilities

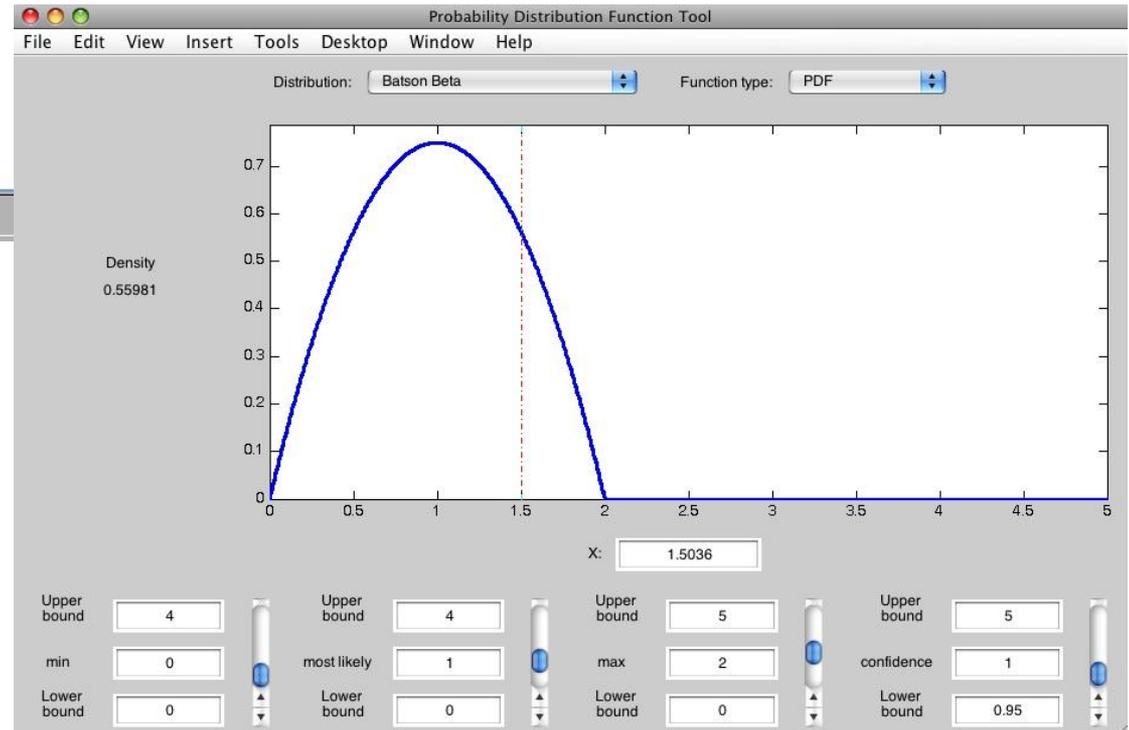
Integration of a collection of tools (each piece designed to attack a specific issue in UQ). The integration represents a unique capability in the field.

- Efficient methods for failure set bounding
  - Optimization-based approach for computing upper bounds on failure domains
- Hybrid methods for efficient estimation of failure probabilities
  - Combining failure set bounding theory with efficient conditional sampling
- First-Order Reliability Method
  - Efficient failure probability approximation for low probability ‘tail’ events
- Efficient deterministic sampling
  - Substantial improvement over conventional Monte Carlo
- Efficient moment propagation methods
  - Useful for propagating trends, e.g., mean & variance of system response
- Probabilistic sensitivity analysis
  - Analyze and rank the relative importance of system parameters
- Response surface tools
  - Radial basis functions and generalized polynomials (with 1<sup>st</sup> and 2<sup>nd</sup> derivatives)

# Graphical User Interface



**Master Parameter List Interface**



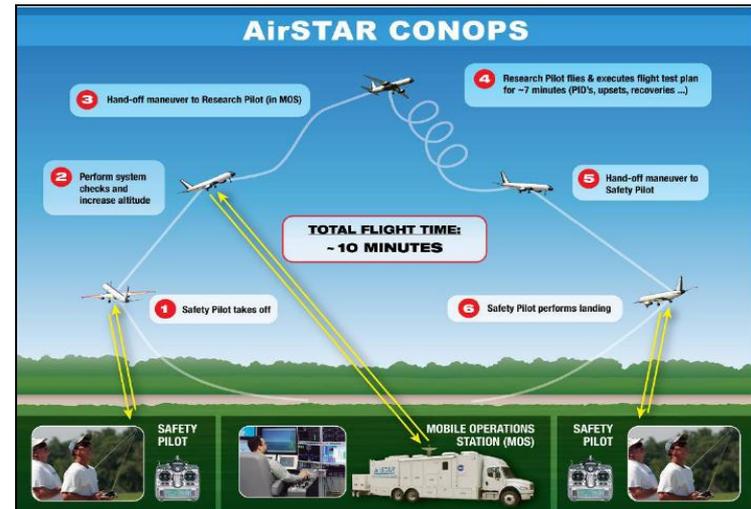
## **Density Function Utility**

- **Currently 20 different distributions supported.**
- **General intervals and deterministic parameters also supported.**

# Example Applications

# Example: GTM Control Analysis & Design

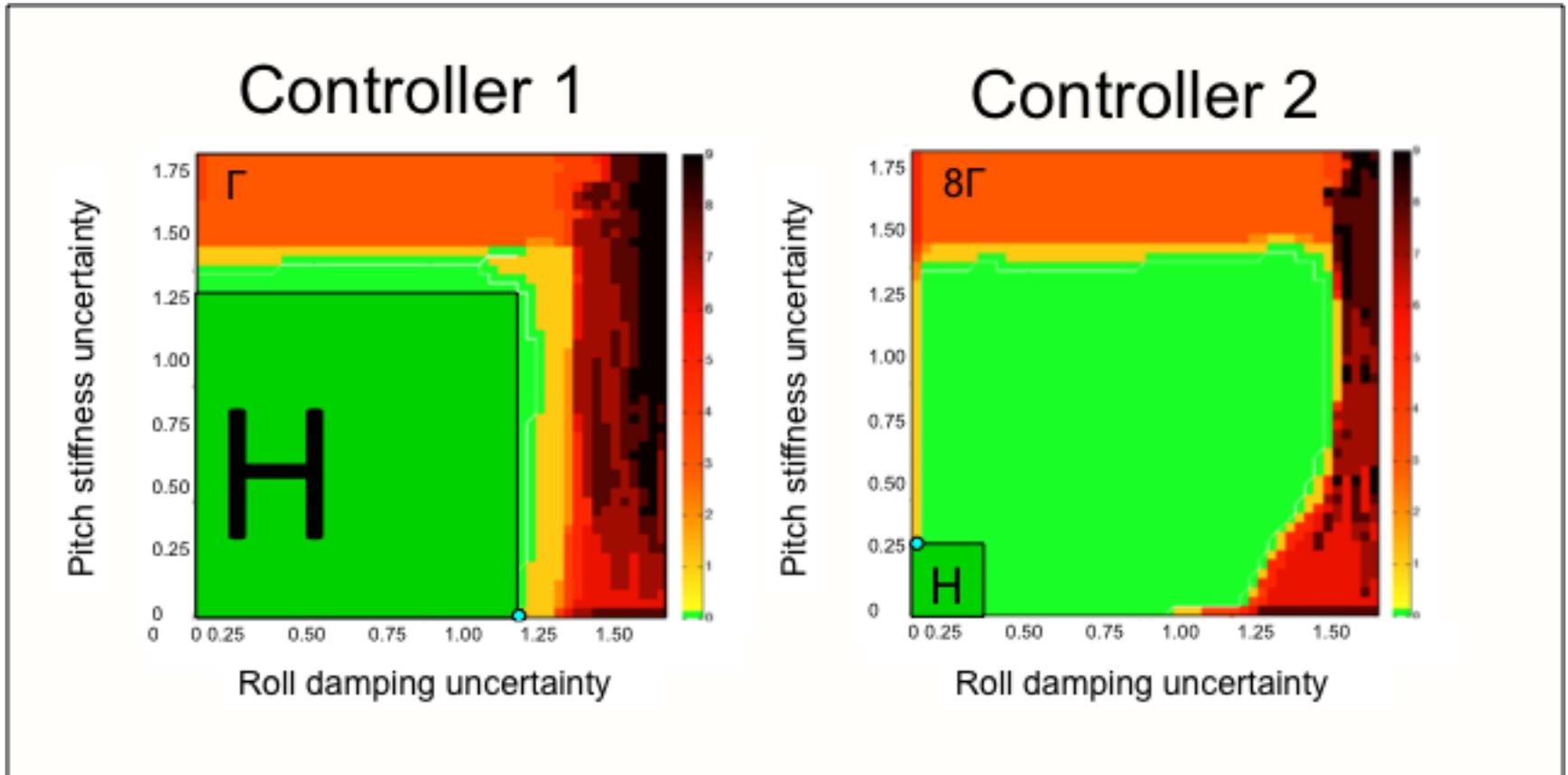
- Subscale, physical model
- High-fidelity Simulink model: non-linear aero, avionics, engine and sensor dynamics, atmospheric model, telemetry effects, time delay, filters, etc. (278 states)



- Control structure: LQR-PI and Model Reference Adaptive
- Uncertain parameters: aerodynamic coefficients
- Requirements: structural integrity, reliable flight envelope, command following, high frequency/residual oscillation

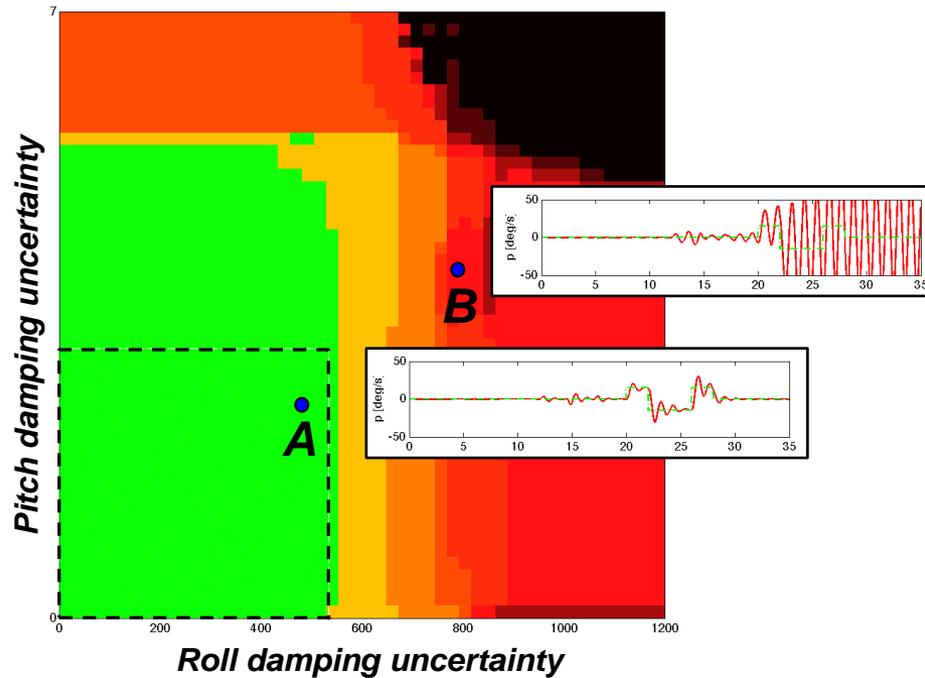
# Homothetic Deformations: Analysis

- Two different controllers analyzed for robustness to aero uncertainties.
- Despite having a larger safe operating region, Controller 2 (high gain controller) has a nearly undetectable failure mode close to nominal point.
- Conventional Monte Carlo is not well-suited to capture this type of failure.

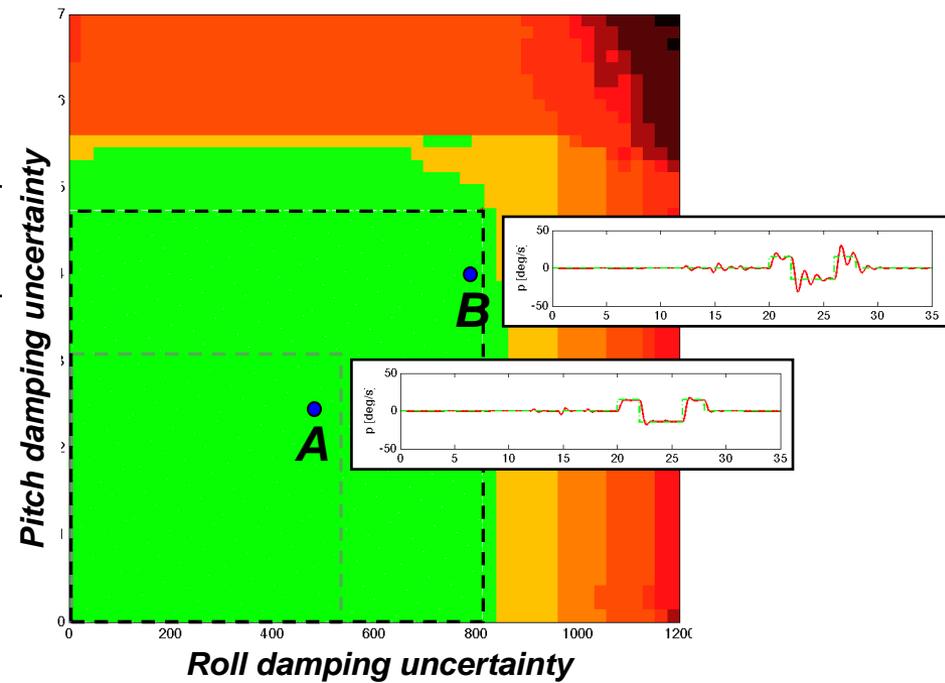


# Homothetic Deformations: Design

*Before Redesign*

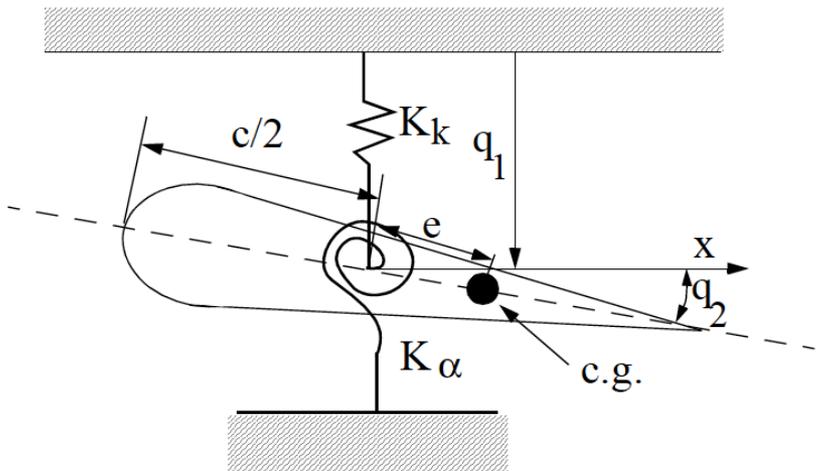


*After Redesign*



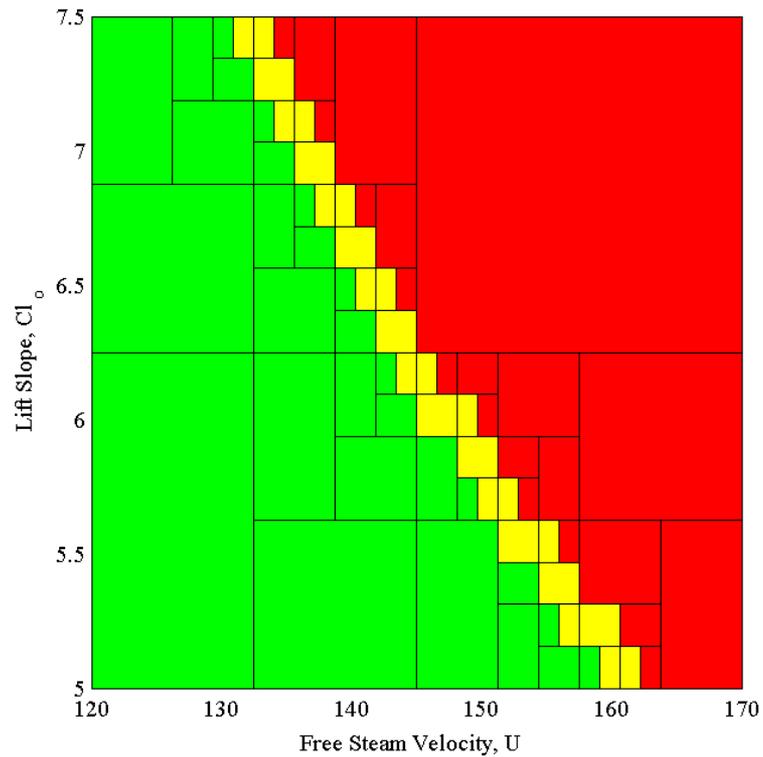
**65 % improvement *in* roll damping uncertainty**

# Simple Aeroelastic Model

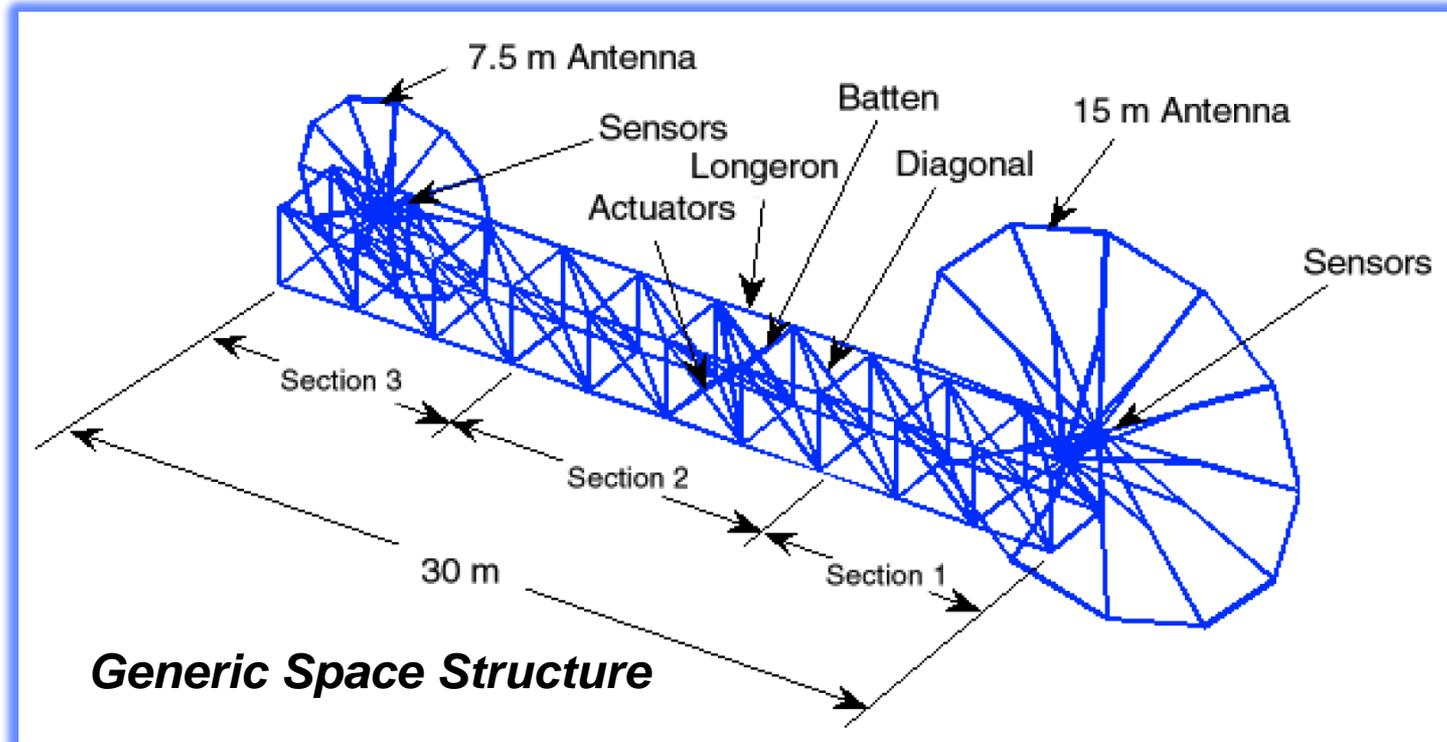


- Linear Aeroelasticity
- 4 uncertain parameters
- Bernstein polynomials
- 5 Routh-Hurwitz constraints

$$A(U, K_k, K_\alpha, Cl_\alpha) = \begin{bmatrix} 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 1 \\ \frac{-10K_k}{9} \left( \frac{-245Cl_\alpha U^2}{288} + \frac{10K_\alpha}{9} \right) & \frac{-245Cl_\alpha U}{288} & 0 & 0 \\ \frac{10K_k}{9} \left( \frac{343Cl_\alpha U^2}{144} - \frac{100K_\alpha}{9} \right) & \frac{343Cl_\alpha U}{144} & 0 & 0 \end{bmatrix}$$

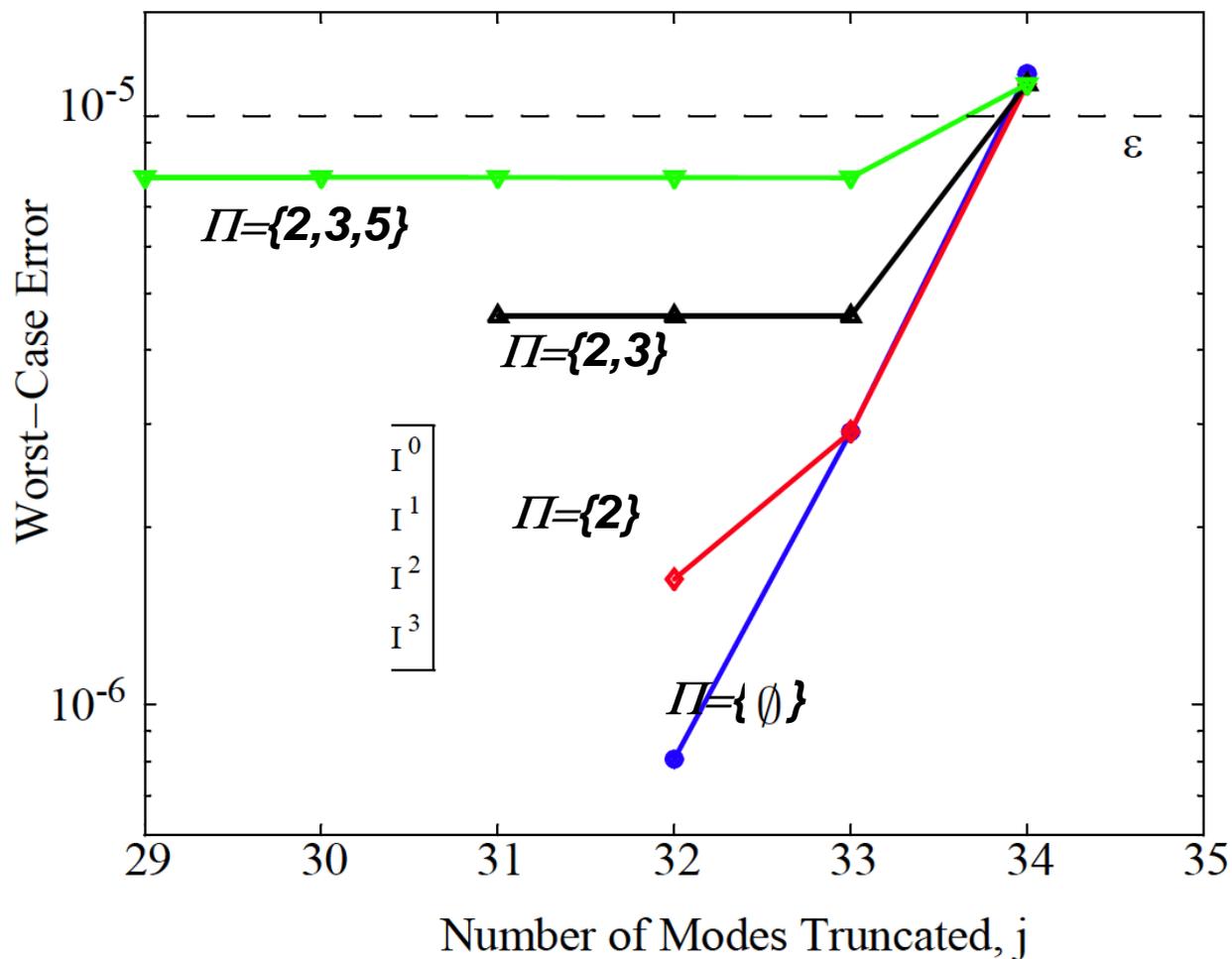


# Example: Flexible Antenna



- Uncertain parameters: outer radius of precision tube members (14 groups, with  $\Delta p = +/- 0.5 \%$ )
- Maximum allowable error,  $\varepsilon=10^{-5}$ , (-100 dB)
- MIMO system (78 states, 3 inputs, 6 outputs)

# Example: Flexible Antenna



# Efforts to Improve State of Practice

- In June 2009, LaRC formed an Uncertainty Based Methods Community Forum (currently 60 members – one AMES CS + one private sector researcher)

*The objective of the UBM Community Forum is to facilitate the cross fertilization of methods, tools, and ideas related to statistical and probabilistic analysis and design for a broad spectrum of engineering applications*

- Foster informal and open discussion of problems in uncertainty based methods
- Discuss success and/or failure
- Share current applications
- Provide a forum for new methods to be vetted

# Conclusions

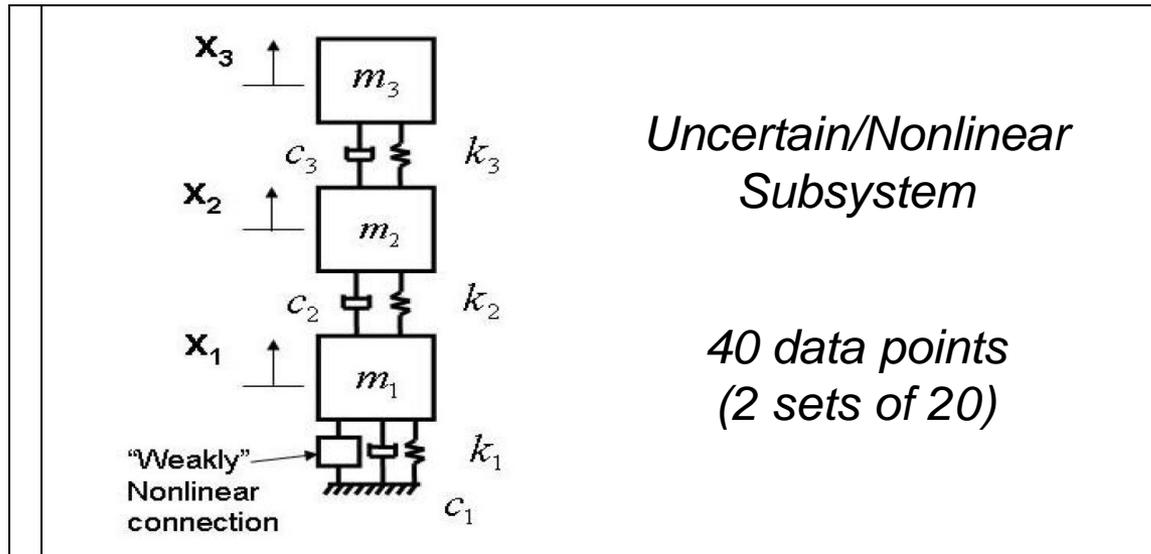
- ***High-fidelity characterization of the failure domain***
- ***Identification of worst-case uncertainty combinations***
- ***Exact failure probability bounds***
- ***Substantially desensitizes the uncertainty analysis from the uncertainty model assumed***

# Validation Challenge Workshop

## Sandia National Laboratory Workshop

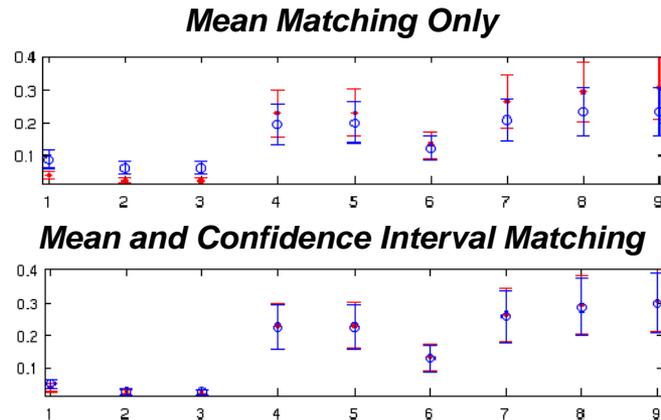
(13 international teams of experts chosen to participate)

**Challenge:** Adequate statistical characterization of uncertainty using limited experimental data.

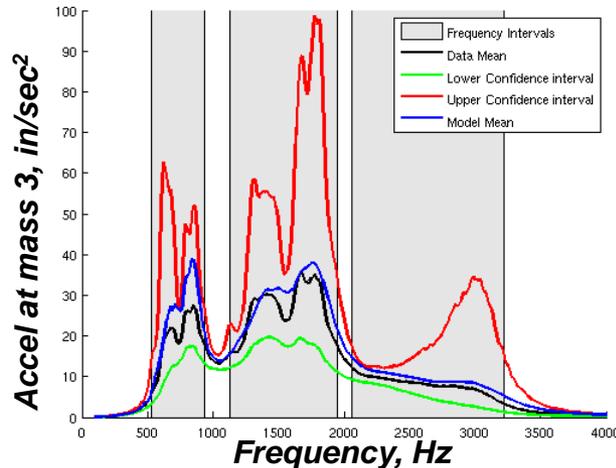


# LaRC's Solution

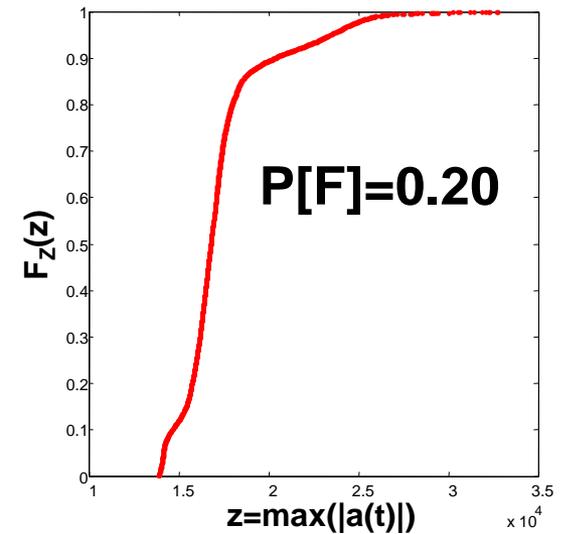
*Statistical matching of parametric uncertainties*



*Statistical response of uncertainty model versus data*



*Results on Target System*



*Regulatory Requirement*  
 $P[z > 18000 \text{ in/sec}^2] < 0.01$