

# **Comparison of Methods for the Propagation of Component Modeling Uncertainty**

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# Outline

- **Motivation**
- **Introduction**
  - **Definitions of Bottom-Up and Top-Down UQ**
- **Development of UQ Procedures via Practical Example**
- **Comparison of Results at Preliminary Level then Following Propagation**
- **Conclusion**





# Motivation

- **Uncertainty Quantification (UQ) – Activity whose goal is to characterize random (and other sources of) variability in a phenomenon**
- **Nondeterministic systems subjected to nondeterministic excitations (and BC, IC, etc.) have nondeterministic responses – Behaviors subject to UQ**
- **Reasons for UQ**
  - **Assessment of system reliability**
  - **Simulation of behaviors and environments**
  - **Validation**
  - **Etc.**

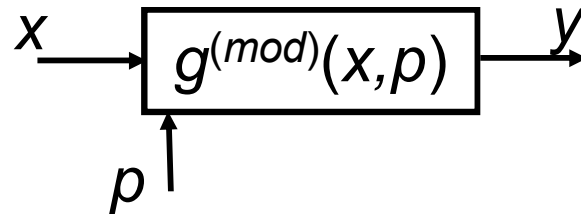
**Whenever there is a UQ requirement, there is a need for techniques to perform UQ**





# Introduction

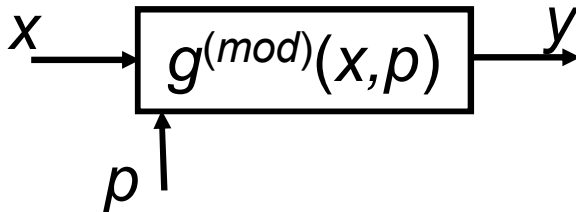
- Probabilistic modeling/analysis - type of UQ to be considered
- Traditional probabilistic analysis usually
  - Starts with probability model of system parameters (excitations, etc.), then
  - Propagates/transforms that information through the model,  $g^{(mod)}(x,p)$ , to probability distribution of measure of system response





# Introduction

- Shortcomings of the traditional probabilistic approach
  - In practical situations (where physical system is being modeled) inadequacy of model form usually ignored
  - Noise on all measurements usually ignored
  - Therefore, probability model of response is incorrect
    - does not capture all correct features of real system response - and usually contains too little variation



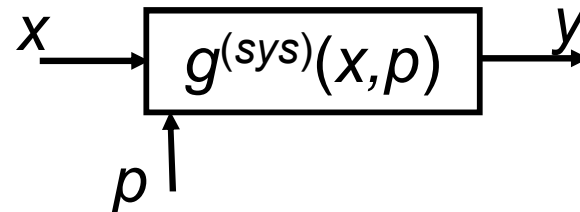
**This approach is known as “Bottom-Up”**





# Introduction

- Alternate approach creates probabilistic model of response, directly (Can be more accurate than bottom-up modeling – It reflects true system behavior)
- (Sometimes variability of parameters is inferred via inversion of  $g^{(mod)}(x,p)$  - not usually)
- Variability of response includes components due to
  - Parameter variability
  - Input variability
  - Measurement noise



**This approach is known as “Top-Down”**





# Introduction

**A comparison is to be performed**

- **Probabilistic behavior of subsystem temporal response is modeled via bottom-up and top-down approaches**
- **Modeling performances compared**
- **Probabilistic models of subsystem propagated through more complex system model to establish probabilistic behavior of temporal response of second system**
- **Performances compared**





# Illustrative Example

- **Project objectives: Create model to predict response behavior of aerospace component and assess stochastic behavior**
- **To test modeling capabilities, consider simple system**
  - Mass supported on three legs that are integral part of base
  - Connection is lap joint - nonlinear
  - Excitation imparted via base
- **Probabilistic behavior of mass response built from probabilistic behavior of joint**





## Illustrative Example (Continued)

- Experiments performed on joint to establish its behavior

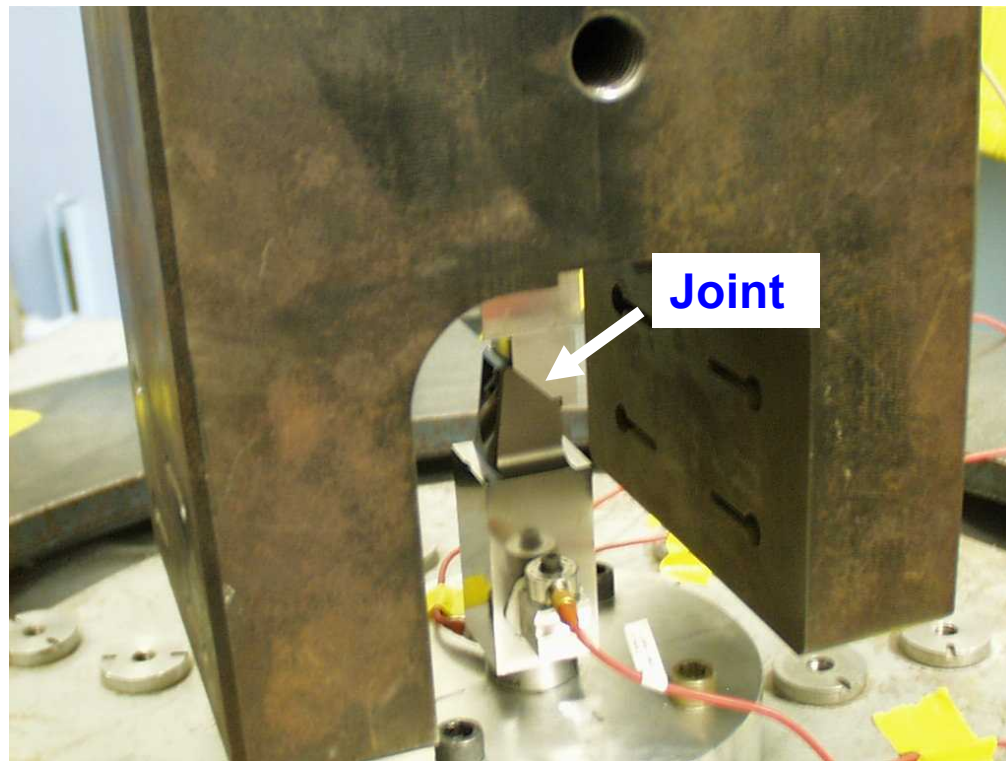


- Connection is lap-joint
- No macro-slip - only micro-slip
- Three nominally identical “tops” and three nominally identical “bottoms” available
- All nine combinations tested
- Test on each combination repeated five times
- Tests excite harmonic motion



## Illustrative Example (Continued)

- Experiments performed on joint to establish its behavior

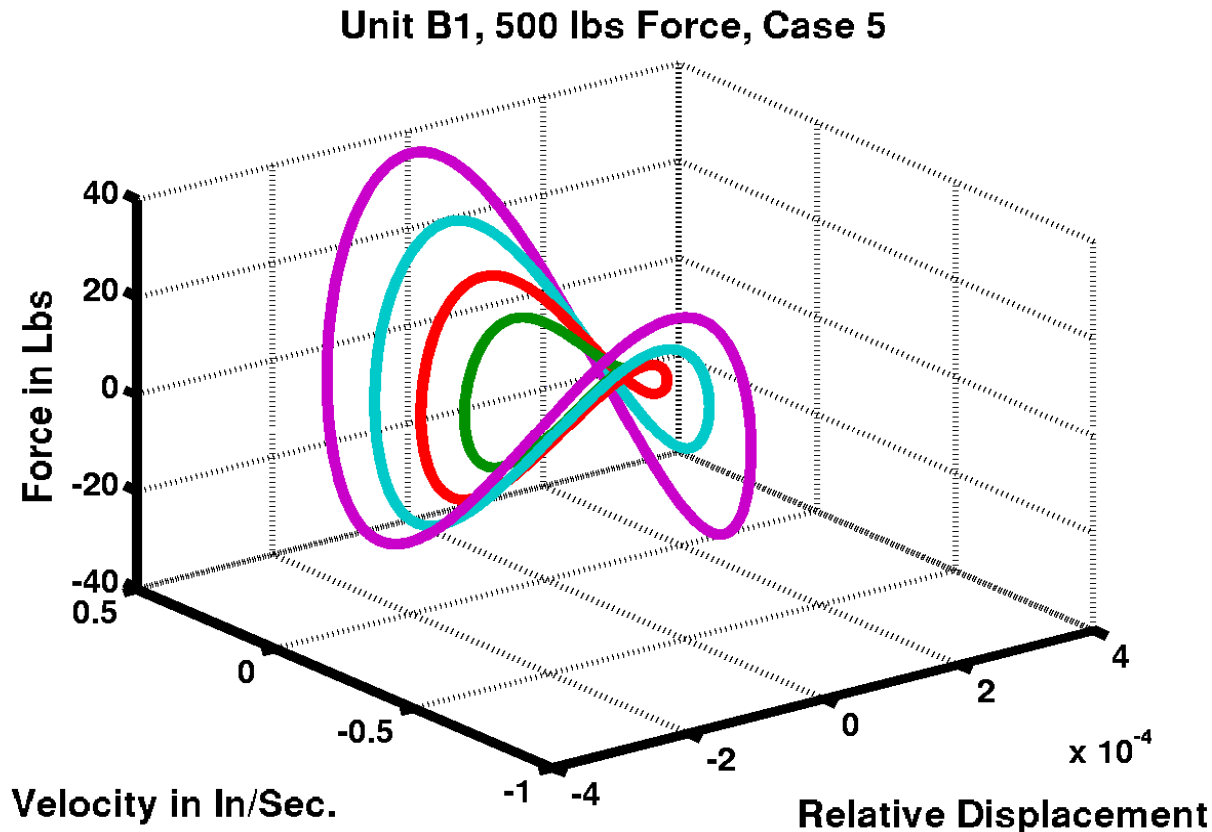


- Excitation generated via electro-dynamic shaker
- Excitation type – swept sine
- Data collected at resonance
- Force across joint measured
- Accelerations at joint-ends measured
- Relative displacement across joint inferred
- System nonlinear – therefore, much force in higher harmonics
- System behaves differently in tension and compression – therefore, fundamental nonlinearity is quadratic



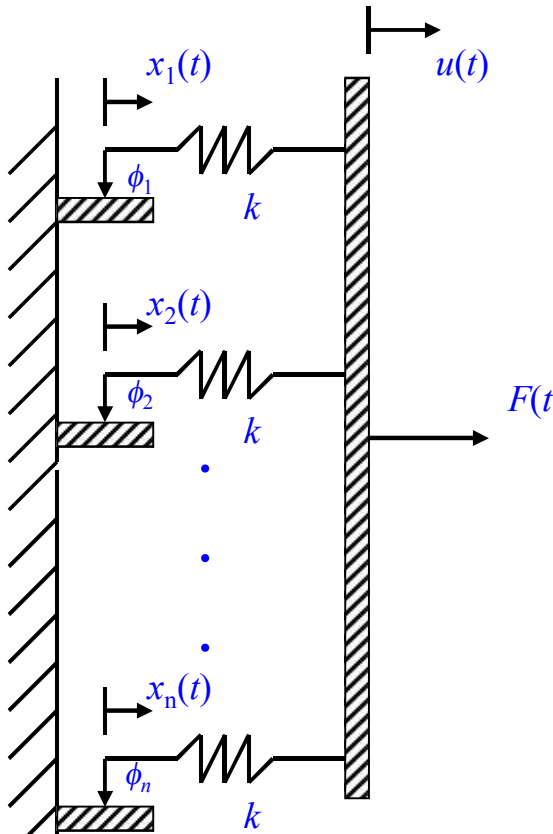
## Illustrative Example (Continued)

- Plot of force versus relative displacement and relative velocity with linear components removed - experiment



## Illustrative Example (Continued)

- Iwan model used for joint behavior



- Model consists of sequence of “Jenkins” elements
- Behavior characterized by stiffnesses and slip displacements
- Hysteretic behavior of Iwan model characterized by distribution of slip displacements in Jenkins elements
- Distribution of slip displacements of sliders is the same in tension and compression
- Therefore, hysteresis loops are anti-symmetric – fundamental nonlinearity is cubic

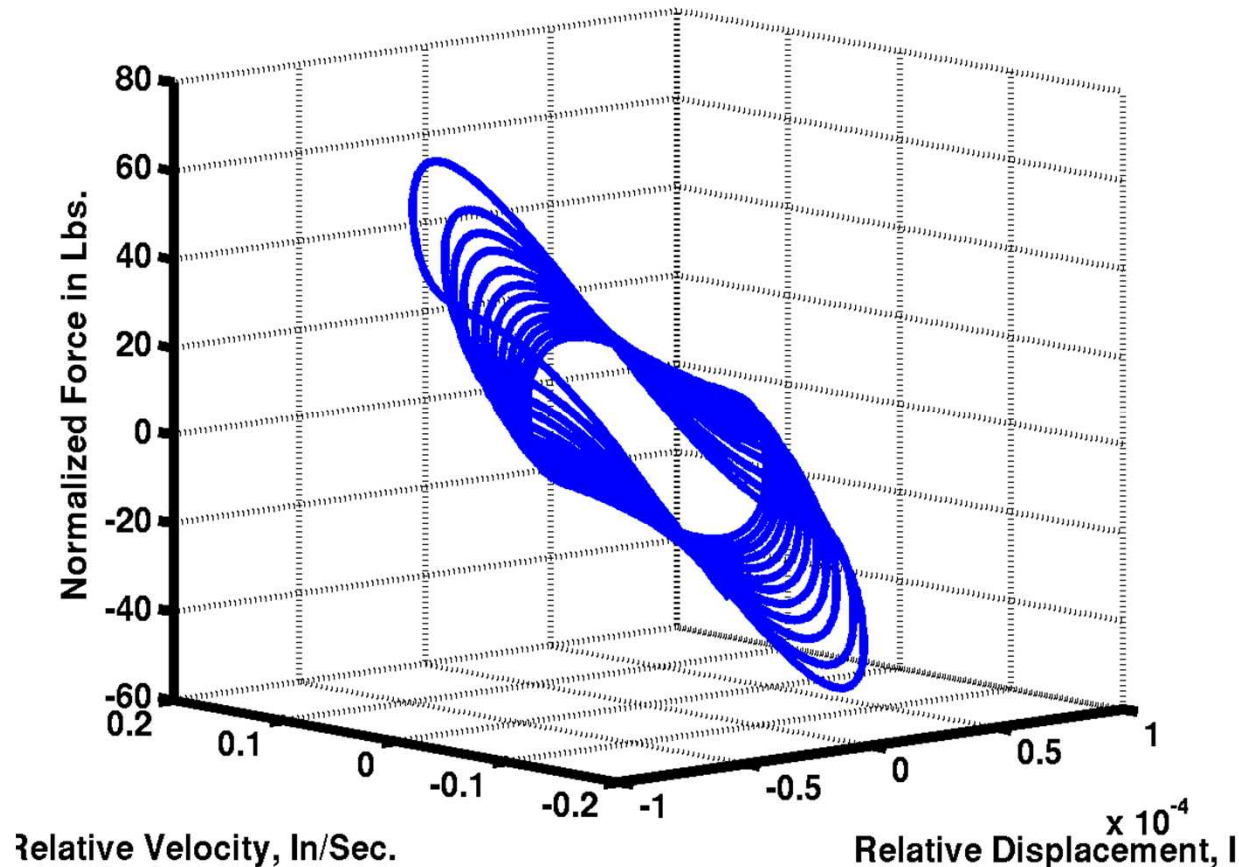




## Illustrative Example (Continued)

- Plot of force versus relative displacement and relative velocity – Iwan model

Iwan Element, Impulse Response





## Illustrative Example (Continued)

- **Top-Down model for time domain response – Fourier series**

$$d^{(test)}(t) \cong \sum_{k=1}^n a_k^{(test)} \cos(k\omega t) + b_k^{(test)} \sin(k\omega t)$$

$$d^{(mod)}(t) \cong \sum_{k=1}^n a_k^{(mod)} \cos(k\omega t) + b_k^{(mod)} \sin(k\omega t)$$

- **Generic covariance matrix of total uncertainty**

$$S_{\tilde{B}\tilde{B}} = \frac{1}{N} \sum_{j=1}^N \{\Delta\tilde{B}\}_j \{\Delta\tilde{B}\}_j^T$$

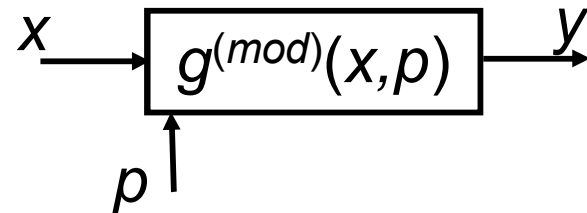
**where  $\{\Delta\tilde{B}\}_j$ ,  $j = 1, \dots, N$ , are vectors of normalized differences between test and model Fourier coefficients**





## Illustrative Example (Continued)

- Bottom-Up model for time domain response



where  $g^{(mod)}(x, p)$  is finite element model of system, including joint modeled as Iwan

- Covariance matrix of Iwan model parameters

$$S_{pp} = \frac{1}{N} \sum_{j=1}^N \{\mathbf{p}\}_j \{\mathbf{p}\}_j^T$$

where  $\mathbf{p}_j, j = 1, \dots, N$ , are vectors of experimentally-derived realizations of Iwan model parameters





## Illustrative Example (Continued)

- Uncertainty of both Bottom-Up and Top-Down models for time domain response can be approximately, linearly propagated to covariances of displacement time histories

$$(S_{dd})_{\tilde{B}} = T_{d\tilde{B}} S_{\tilde{B}\tilde{B}} T_{d\tilde{B}}^T$$

$$(S_{dd})_p = T_{dp} S_{pp} T_{dp}^T$$

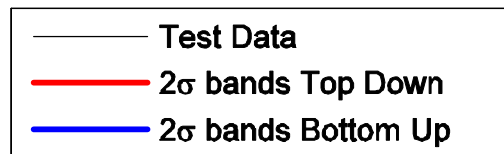
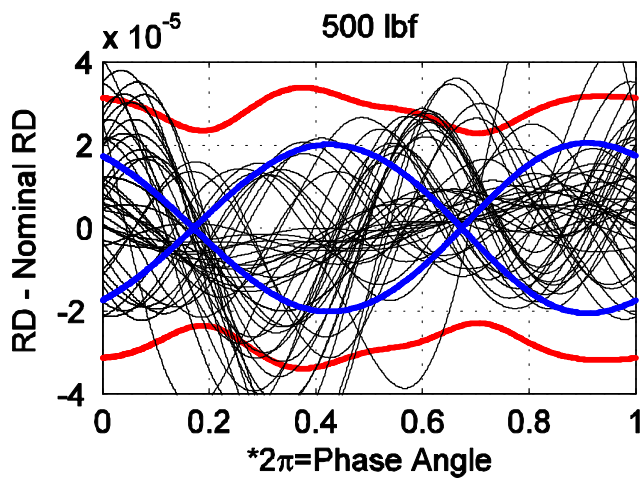
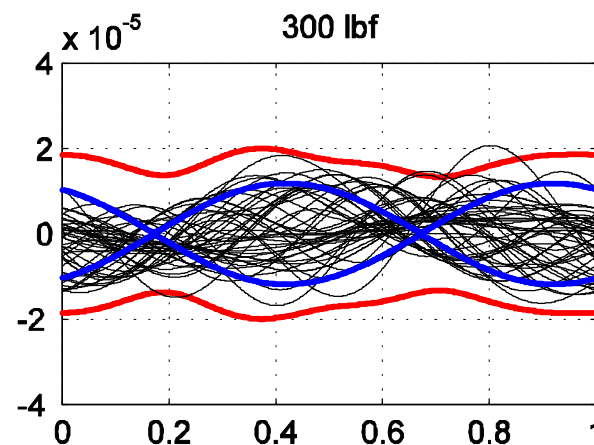
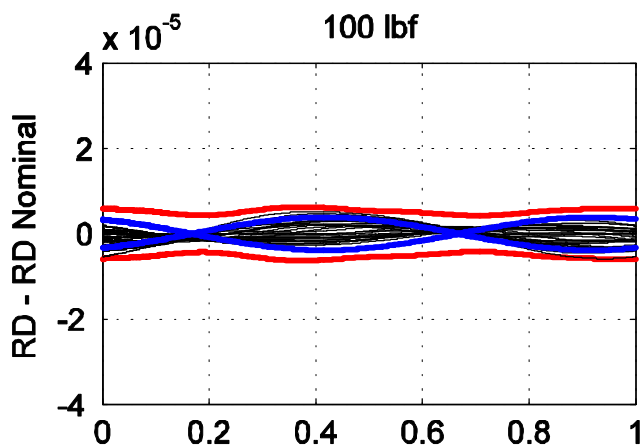
where  $T_{d\tilde{B}}$  is the sensitivity matrix relating normalized Fourier coefficients to displacement time history, and  $T_{dp}$  is the sensitivity relating Iwan parameters to displacement time histories





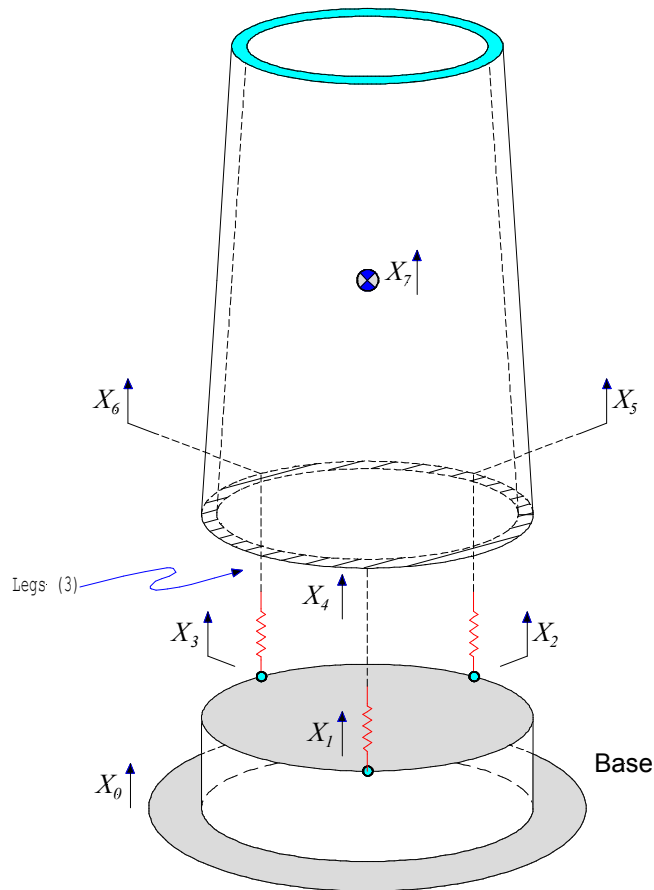
# Illustrative Example (Continued)

- Result



## Illustrative Example (Continued)

- Next step – Propagate uncertainty of joint element to uncertainty of three-legged system

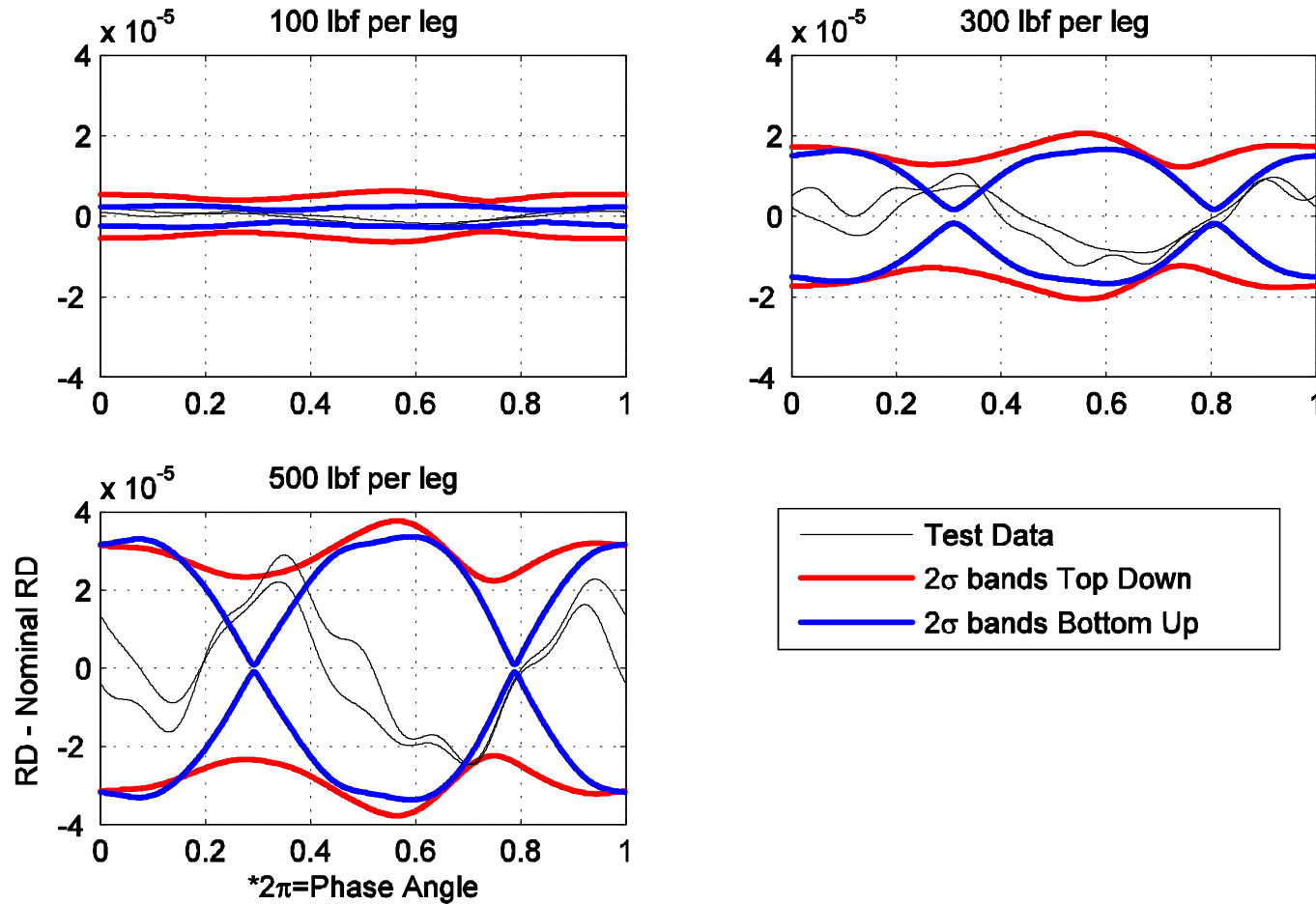


How is this accomplished?

- In this case, uncertainty associated with individual leg deformations propagated to center-of-gravity motion
- Mass assumed rigid
- Various assumptions made regarding correlations among leg motions

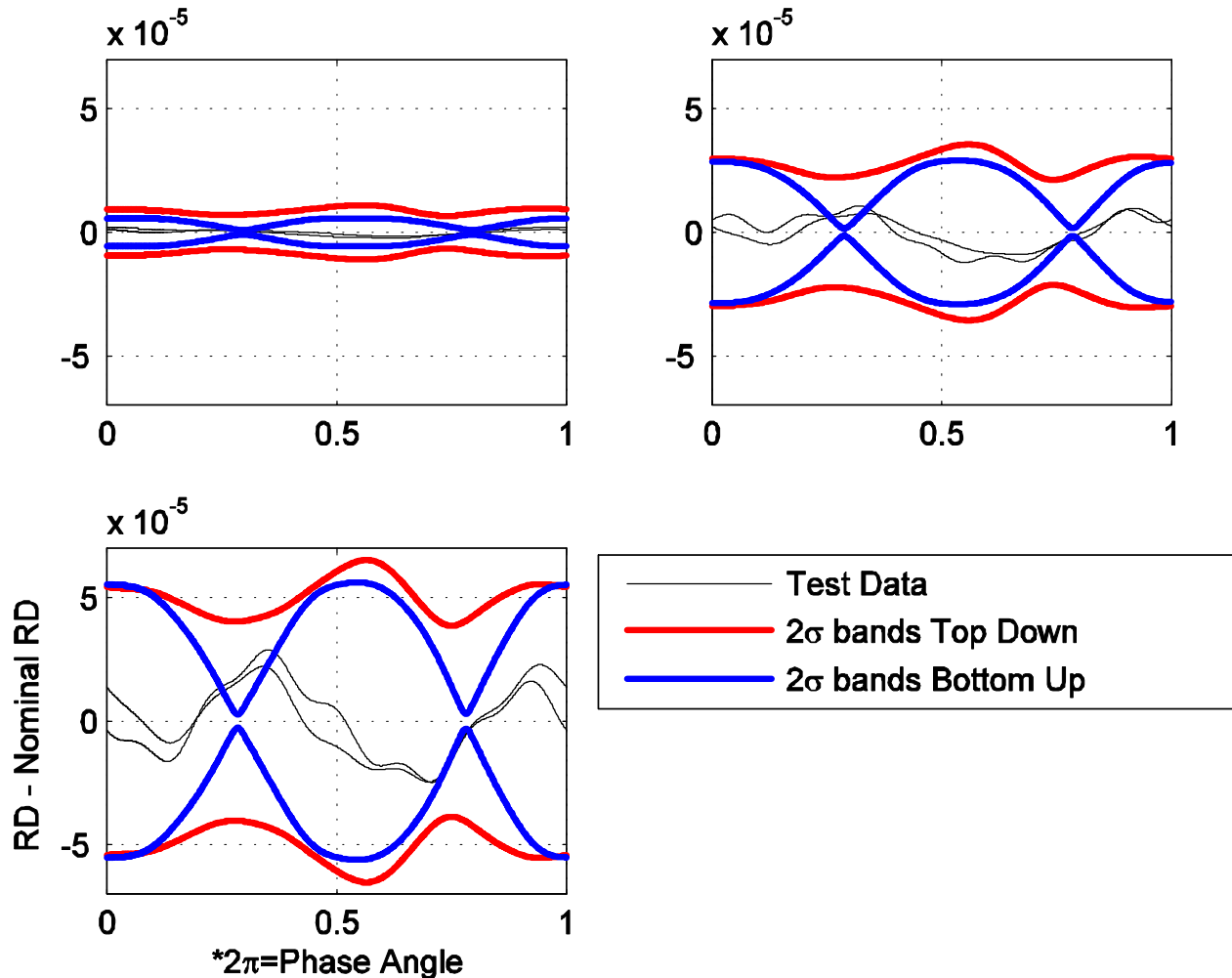
## Illustrative Example (Continued)

- Results – Uncorrelated single-leg UQ parameters



## Illustrative Example (Continued)

- Results – Perfectly correlated single-leg UQ parameters





# Conclusions

- **Bottom-Up and Top-Down UQ analysis approaches compared**
- **Top-Down approach**
  - Shown to more accurately reflect response variability – form and level – for time history of response
  - Propagations use linear form – rely on sensitivity and covariance information, direct to perform
- **Bottom-Up approach**
  - Provides framework for relating system or component parameter variability to response variability
  - Relies on system model, at least, for first propagation, therefore, less accurate than Top-down



