

Validation Methodology for Passenger Safety

*Development of a Verified and Validated, Parametric
Cervical Spine Injury Model*

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San Antonio, TX



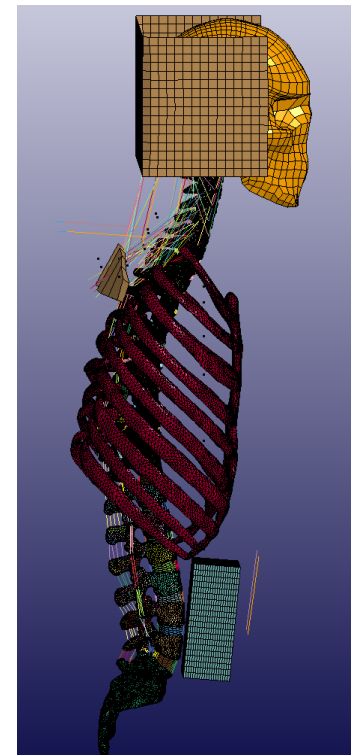
Motivation and Support

- Project sponsored by the Naval Air Warfare Center - Aircraft Division, Dr. Barry Shender, PM
- Assess probability of injury for military pilots in severe situations
 - Ejections and carrier landings
 - Female aviators in cockpit systems designed for males
 - Added mass in helmet systems
 - Long-term high-g exposures



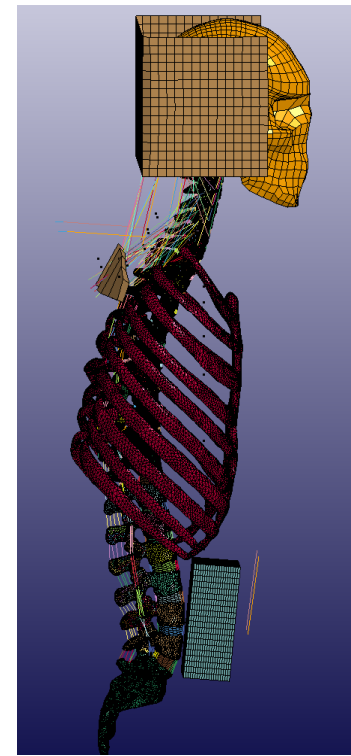
Program Goals

- Develop a predictive model of the musculoskeletal system
- Employ hierarchical model validation methodology
- Investigate sex and weight effects on probability of cervical spine injury
- Develop a tool to assist with designing new helmet systems, seats, etc.



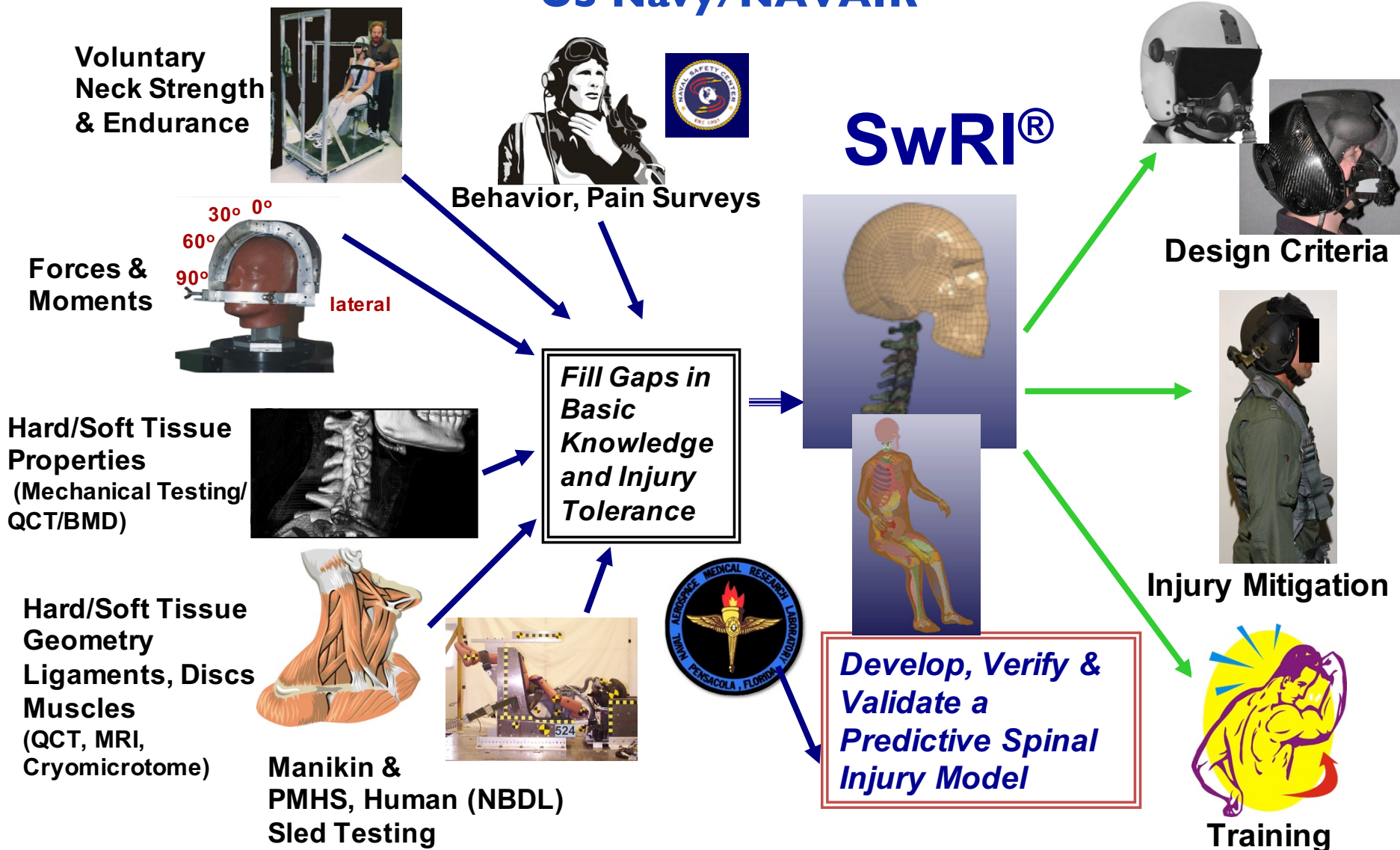
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Multidisciplinary Team Approach

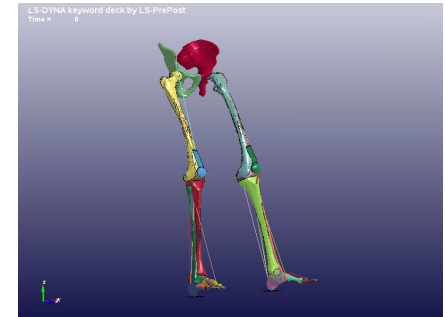
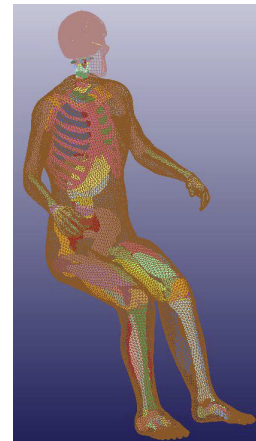
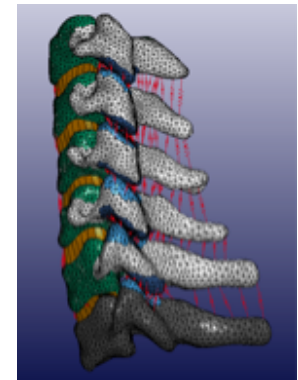
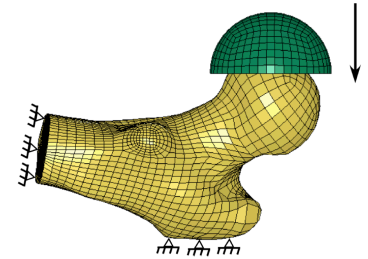
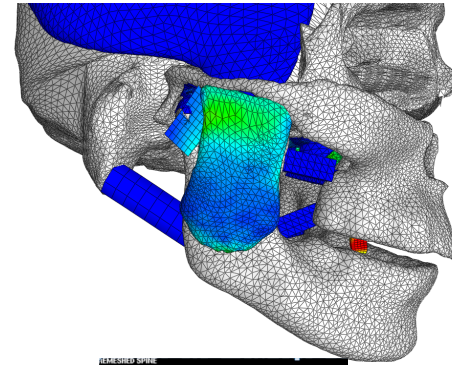
US Navy/NAVAIR



ADVANCED SCIENCE. APPLIED TECHNOLOGY.

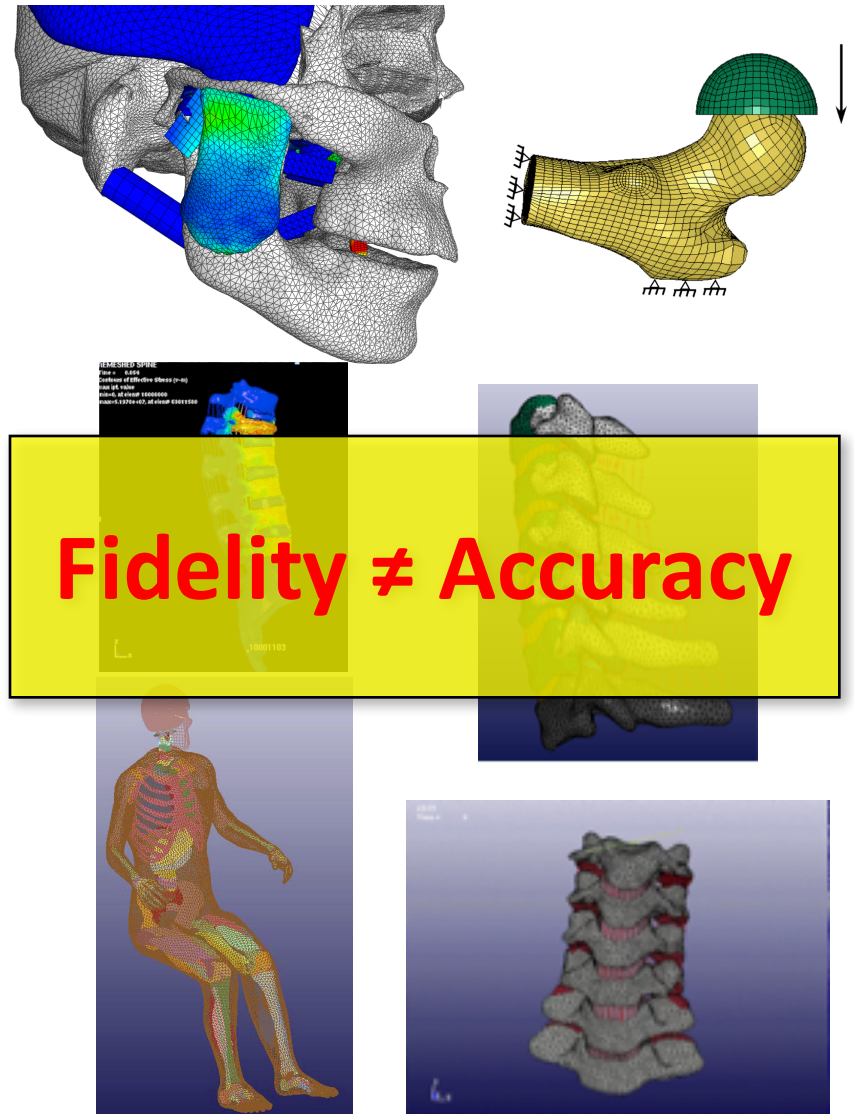
Computational Modeling for Biomechanical Analysis

- Relatively easy to construct high fidelity models from high quality 3D image data
 - powerful geometry modeling and meshing software
 - high performance computational resources
 - Resulting models “looks” almost identical to the actual biological system
- Non-linear material constitutive models with properties either derived from experimental data or reported in the literature
- Large deformation, and motion defined by sliding contact between complex, deformable articulating surfaces
- Result: High Fidelity Computational Models



Model Verification and Validation (V&V)

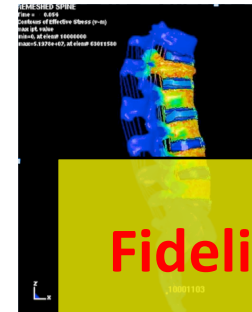
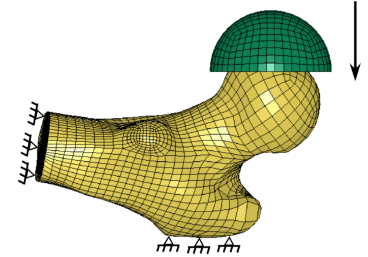
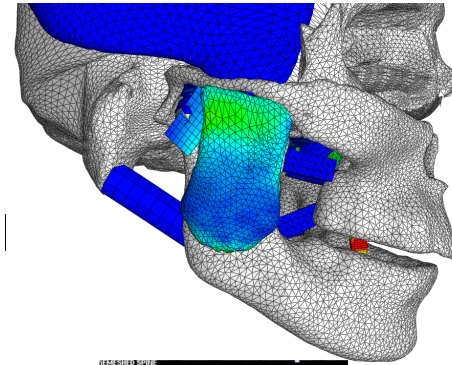
- High fidelity should not be confused with model credibility
- High fidelity is necessary but not sufficient
 - Fidelity is the result of modeling tools (pre-processor, FE code, etc.) computational speed, etc.
- Model credibility is the result of specific and rigorous model V&V



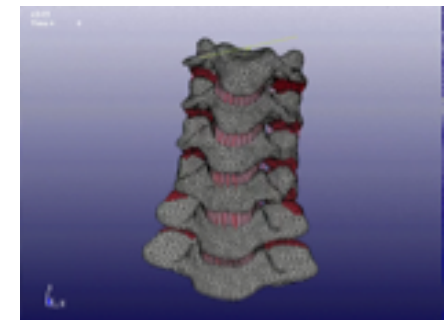
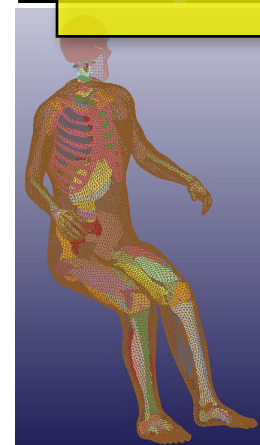
Introduction

Why Model Verification and Validation (V&V)

- Fidelity does not mean accuracy
- Decision makers want to know:
 - What is the **error** between the model and tests?
 - How much **confidence** do we have in the model predictions?
 - Can we use these models to predict occupant injury?
 - Can we design safer systems using these models?
 - How accurate are these models for decision making?
- Model Verification and Validation can help answer these questions



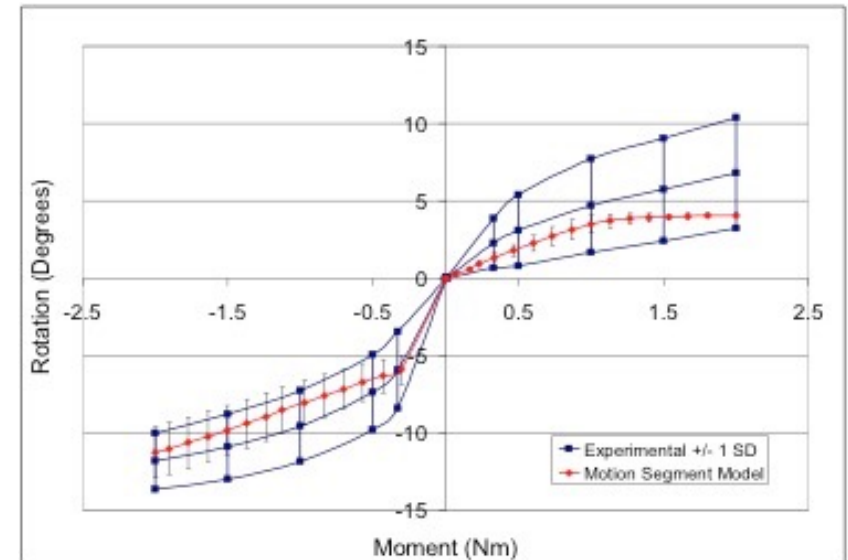
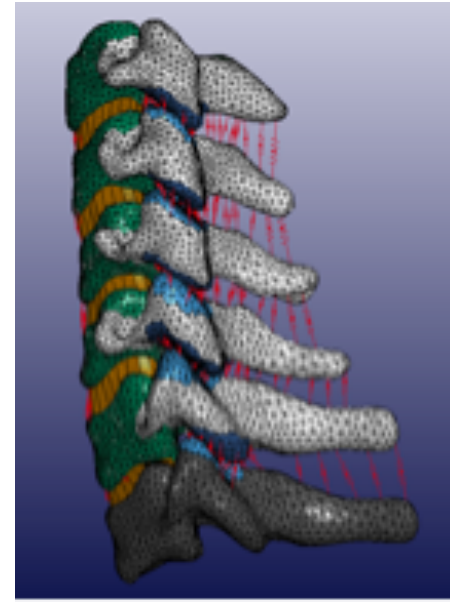
Fidelity ≠ Accuracy



Model Validation Example

Traditional Approach

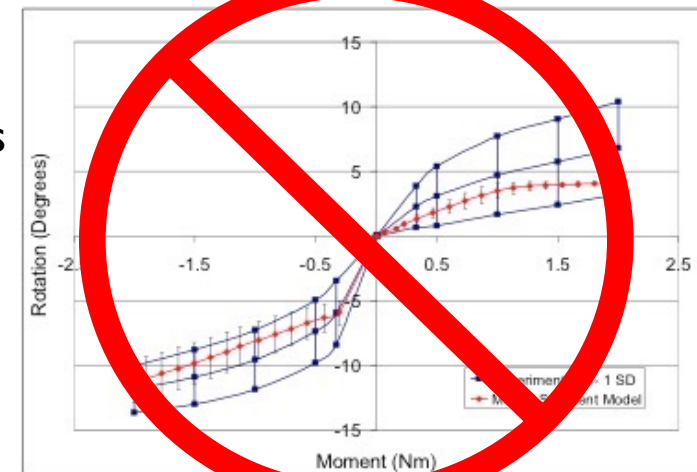
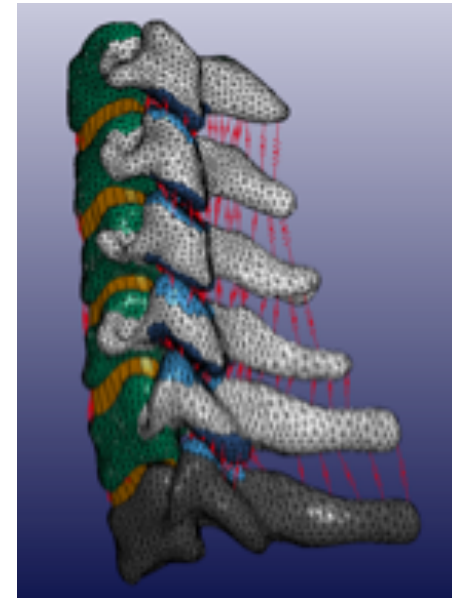
- Construct FE model of target system
- Use material property data from the literature
- Apply boundary conditions to simulate experiment found in the literature
- Compare (overlay plot) of model predicted response with experiment results
- Validation:
 - model is valid if prediction falls within experimental corridors



Model Validation

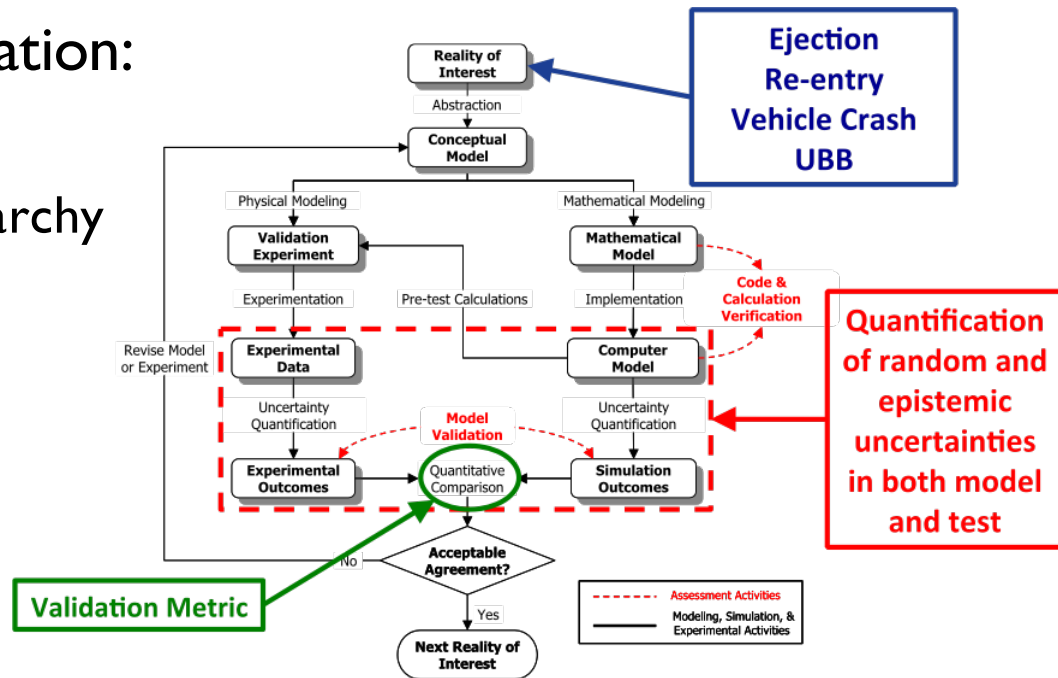
Issues with Traditional Approach

- Details of the experiment are often not well known or understood by the modeling team
 - Experiment boundary conditions
 - Size and shape of exp. specimen
 - Can the experiment be modeled?
- Material properties often “tuned” or selected to match high level structural response
 - Range of values in the literature - let’s pick the values that give us a good match
 - Right answer for the wrong reason
- Corridor limits are arbitrary (± 1 SD)
- Reducing the quality of the experimental data improves the chance that the model is valid (not good!)
- **Mismatch not quantified**
- **How credible are these models for decision making?**



Validation Process

- The validation process has the goal of assessing the predictive capability of the model by quantitatively comparing the predictive results of the model with validation experiments.
- Three key elements of Validation:
 - Validation Experiments
 - Defined by validation hierarchy
 - Uncertainty Quantification
 - Experiment
 - Model
 - Validation Metrics
 - Quantification of error

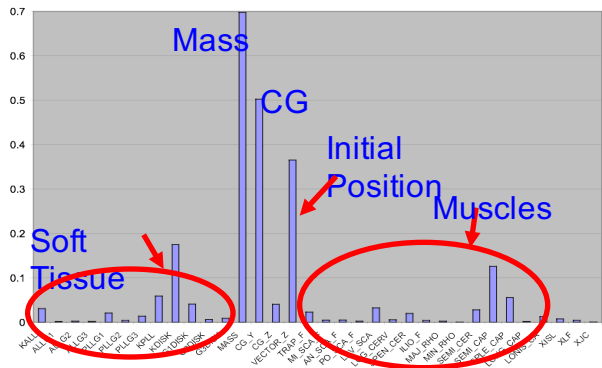
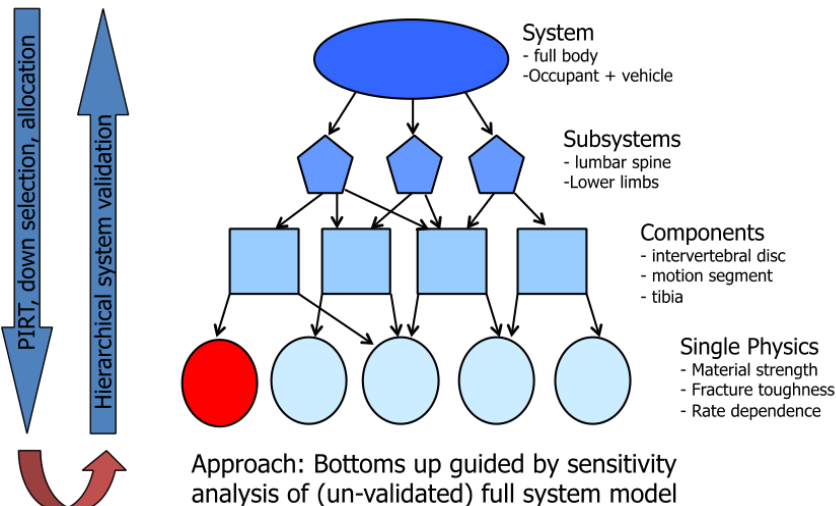


Approach based on ASME V&V 10-2006 "Guide for V&V in Computational Solid Mechanics"

Hierarchical Model V&V Approach

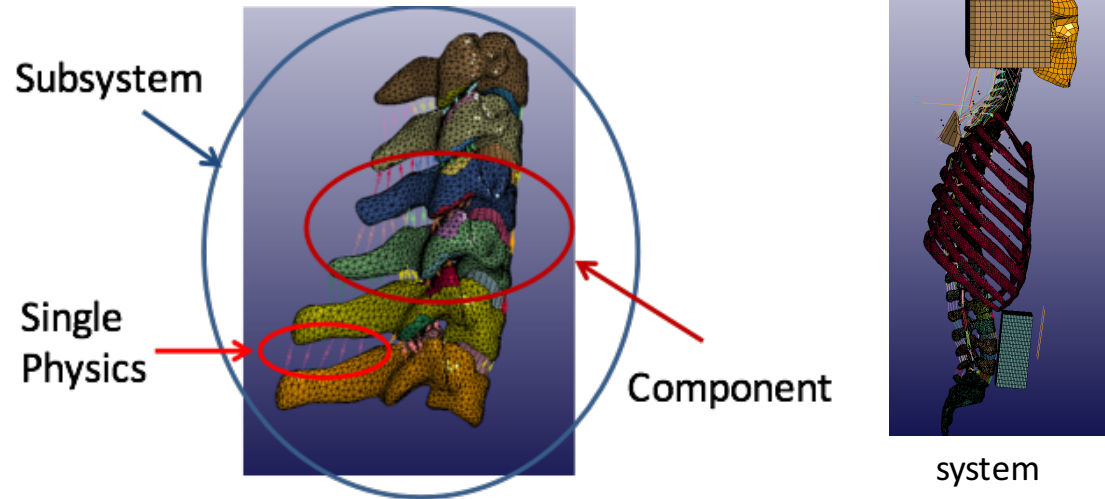
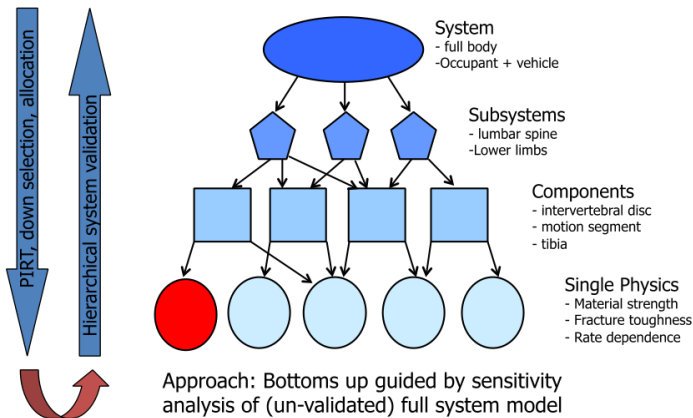
ASME V&V-10 Guidelines

- Customer/stakeholder establishes intended use and top-level validation requirement
- Validation hierarchy
 - Breaks the problem into smaller parts
 - Validation process employed for every element in the hierarchy (ideally)
 - Allows the model to be challenged (and proven) step by step
 - Dramatically increases likelihood of right answer for the right reason
- Validation team constructs hierarchy, establishes sub-level metrics and validation requirements
 - Modeling and experiment teams work closely together to define hierarchy and experiments/simulations
 - Experiments are designed expressly for model validation
- In general, validation requirements will be increasingly more stringent in lower levels
- Single physics and components used to identify material parameters
- Material models “locked-in” at more complex system levels
- Full system (un-validated) sensitivity analysis can provide guidance



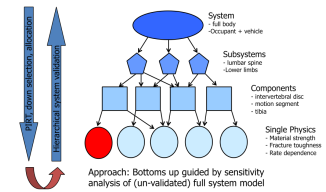
Hierarchical Model V&V Approach

- **Levels 1 & 2 : Material model parameter identification - fit to experimental data**
- Levels 3 & 4: No model fitting/tweaking/calibrating
 - Model performance validated against independent set of experiments
- Supporting experiments performed at:
 - Medical College of Wisconsin – Pintar and Yoganandan
 - University of Virginia/Duke University – Bass and Lucas



Hierarchical Model V&V

Single Physics: Material Model Parameter Identification

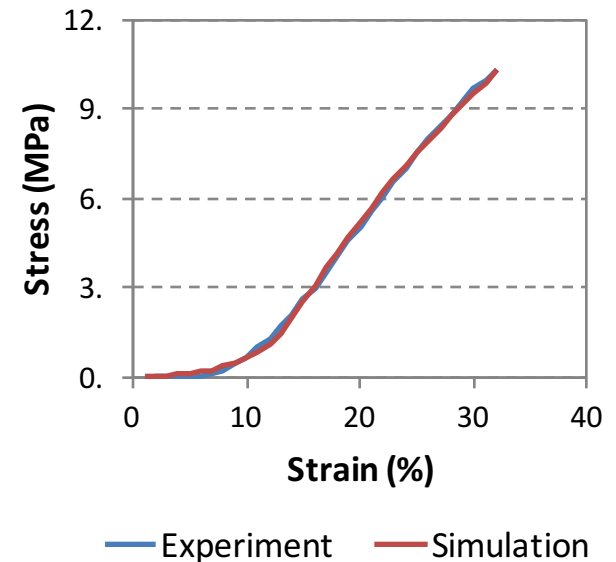
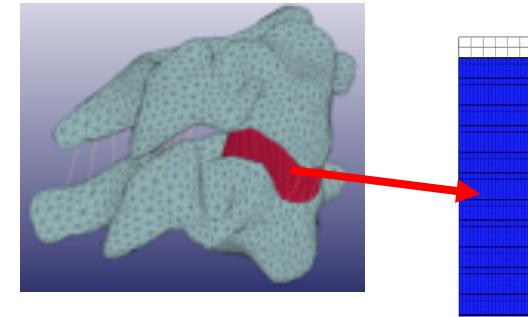


Experimental setup

- Isolate individual intervertebral disc lamellae
- Loaded in tension until failure
- Resulting stress-strain relationship recorded
- 67 experiments performed
- Experiments performed by Stemper, Yoganandan, Pintar

Simulation optimization

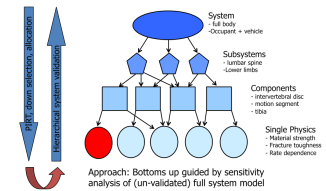
- Isolated annulus tissue model
- Replicated experiment boundary conditions
- LS-DYNA transversely isotropic quasi-linear viscoelastic constitutive model
- Non-linear least squares optimization for each experiment
- Target is force-time history
- Statistical distribution of parameters determined



Parameter	Average	StDev
c1	2.97e5	2.92e5
c3	3.13e4	2.60e4
c4	25.7	8.1
c5	1.80e7	1.16e7
λ	1.14	0.05

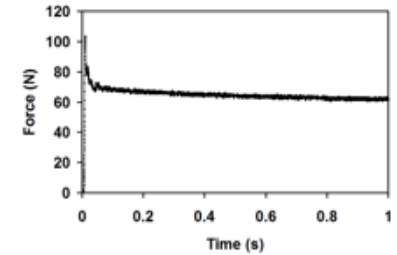
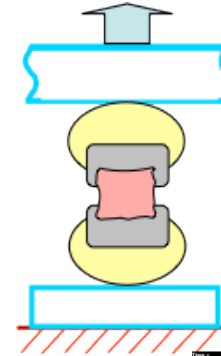
Hierarchical Model V&V

Single Physics: Material Model Parameter Identification



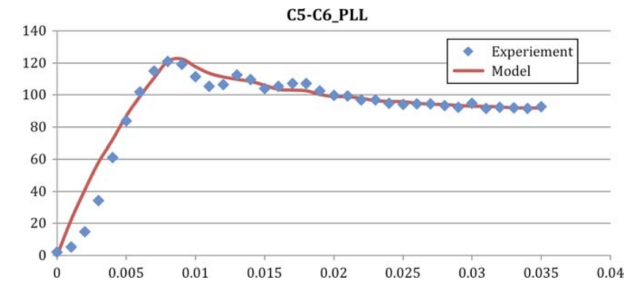
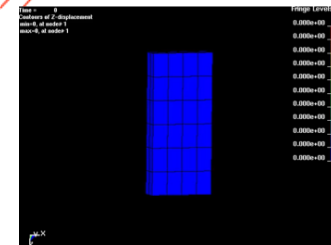
• Experimental Setup

- Dynamic relaxation experiments on isolated ligaments from 6 male and 6 female cadavers (ALL, PLL)
- 25% strain input
- Held for 1 minute - force relaxation recorded
- Scott Lucas and Dale Bass – UVa



• Simulation

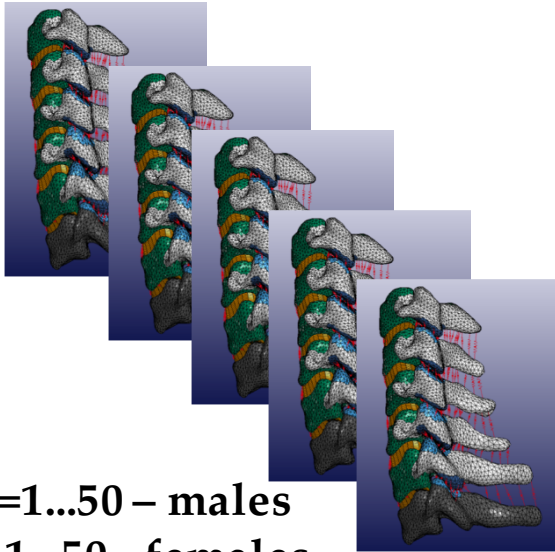
- Isolated ligament model
- Replicated experiment boundary conditions
- LS-DYNA transversely isotropic quasi-linear viscoelastic constitutive model
- Non-linear least squares optimization
- Target is force-time history
- Statistical distribution of material model parameters determined



[1] W. L. Francis, T. D. Eliason, Ben H Thacker, G. R. Paskoff, B. S. Shender, and D. P. Nicoletta, "Implementation and validation of probabilistic models of the anterior longitudinal ligament and posterior longitudinal ligament of the cervical spine.," *Comput Methods Biomech Biomed Engin*, pp. –, Oct. 2012.

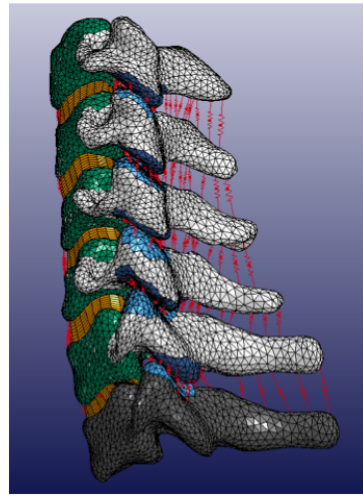
Accounting for Anatomic Variability

Statistical Shape and Trait Analysis



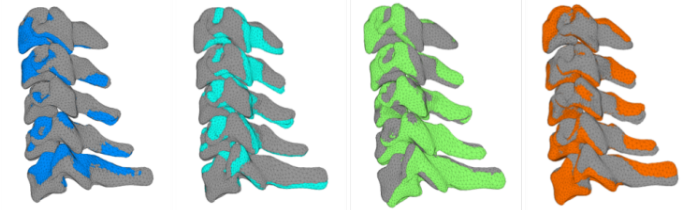
$i=1...50$ – males
 $i=1...50$ - females

$p_i = (v_{1x}, v_{1y}, v_{1z}, \dots, v_{jx}, v_{jy}, v_{jz})^T$
QCT Scans of 100 Individuals
Individual Models



$$\bar{p} = \frac{1}{n} \sum_{i=1}^n p_i$$

Average Model



$$p_v = \bar{p} + \sum_{j=1}^m c_j \sqrt{\lambda_j} q_j$$

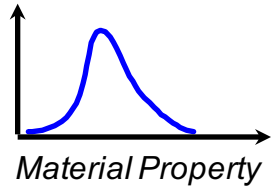
Parametric SSM Model

- Parametric Model = Average Model + **weighting factors** x Principal Components
- Weighting factors contain all variability within the population of interest
- Compact and efficient representation of complex anatomy
- Represent >95% of population anatomic variability with less than 10 variables

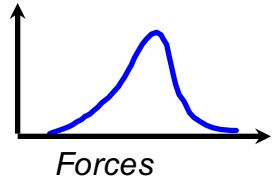
Characterizing Uncertainties

Probabilistic Computational Model

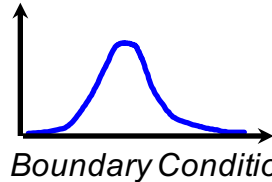
Aleatory uncertainty
Operational Conditions
Population Group



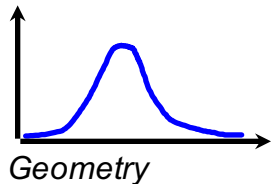
COV's~100%



unknown & variable

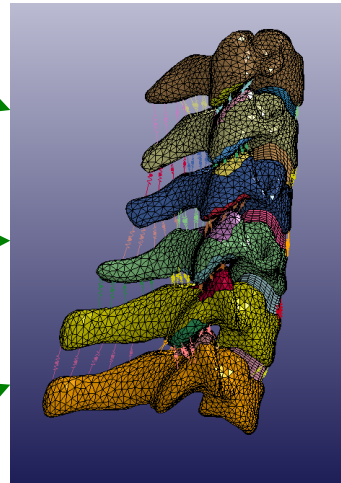


unknown & variable

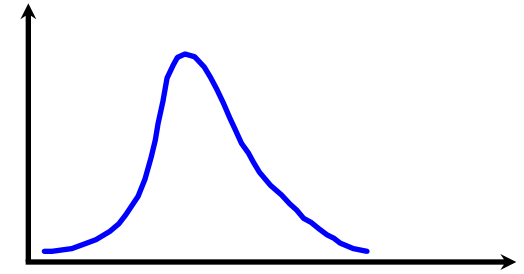


COV's~20%

Finite Element
Model



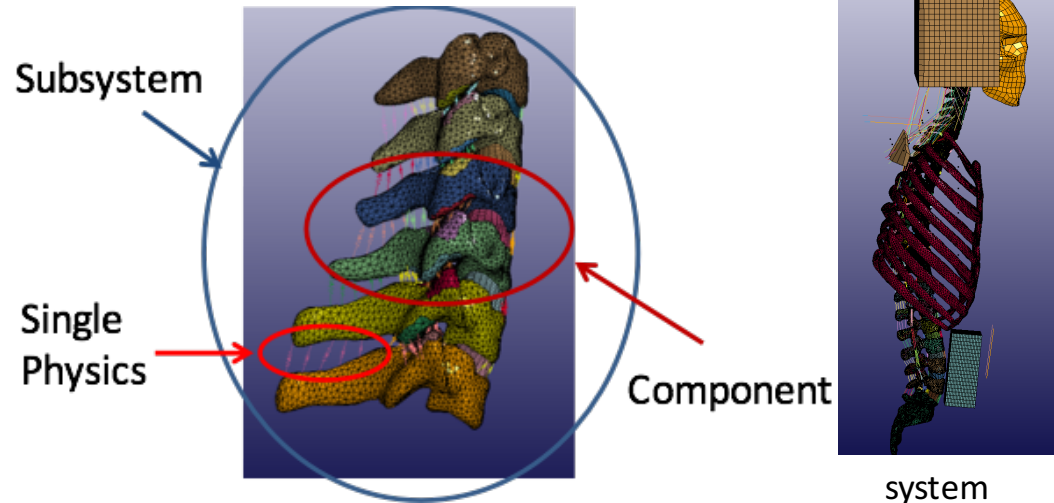
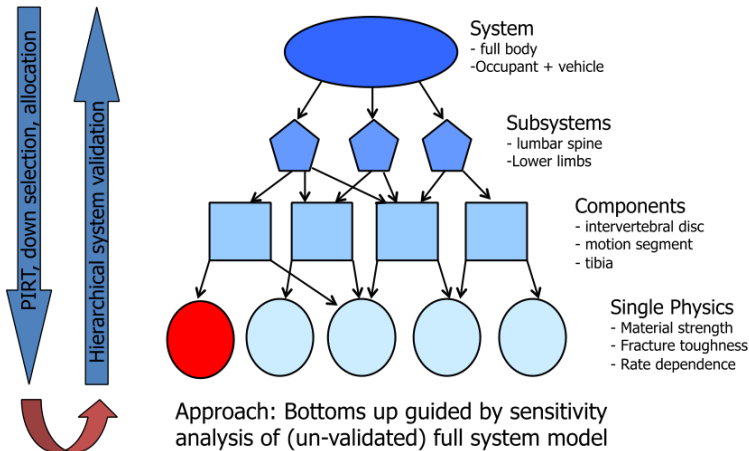
Predicted
Probabilistic
Response



Model Response
Forces, Stress, Strain, Deflections,
etc.

Hierarchical Model V&V Approach

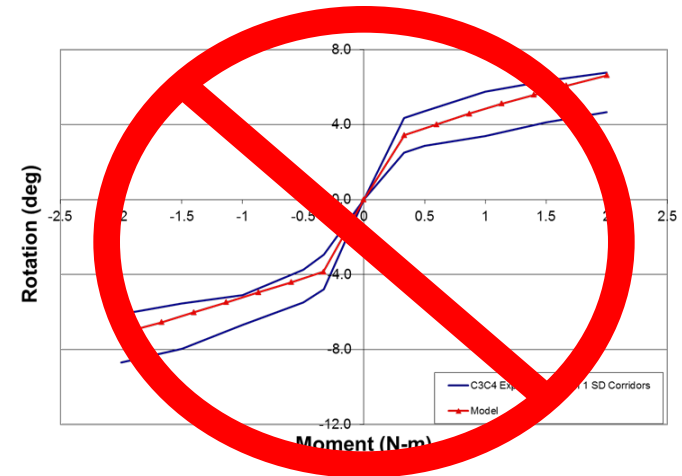
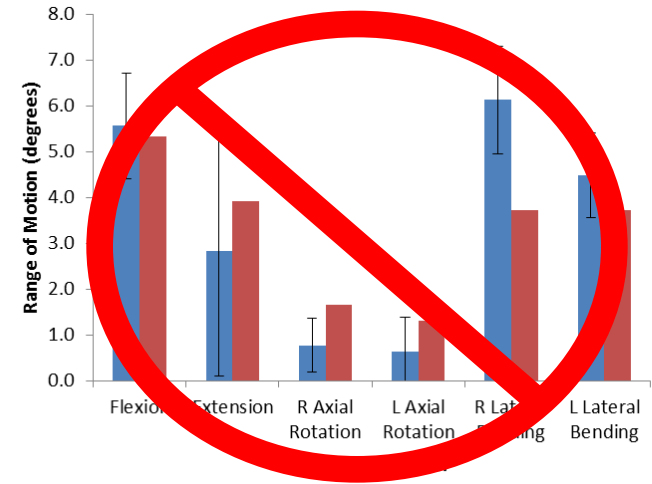
- Levels 1 & 2 : Material model parameter identification - fit to experimental data
- **Levels 3 & 4: No model fitting/tweaking/calibrating**
 - Model performance **validated** against independent set of experiments



Validation Metrics

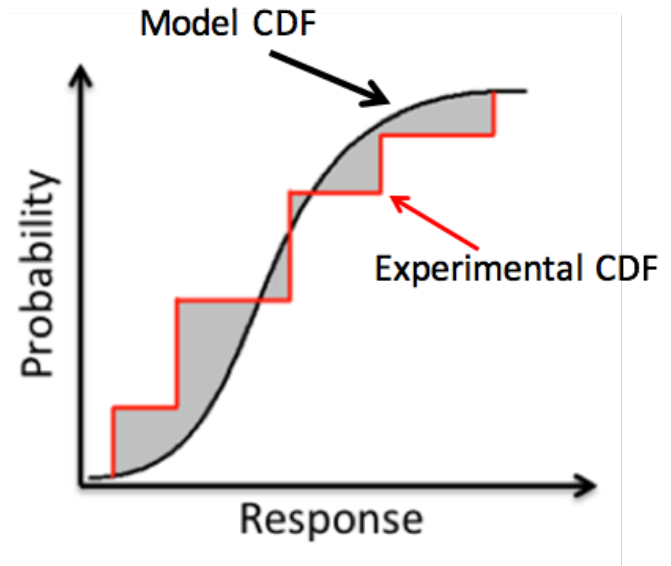
How do you define valid?

- A metric is the quantitative *measure* of the mismatch between model predictions and experimental data
- Typically some type of a difference measure in system response quantities (statistics, probability distributions, etc.)
- Desired features of a validation metric
 - **Consider uncertainties in both the model and the experiment – implies a statistical comparison**
 - Reflect only the comparison (not the adequacy)



Probabilistic Validation - Area Metric

- Calculates the area between the experimental CDF and predicted model CDF
 - Compares mean response and variability between prediction and experiment
 - Gives quantitative measure of model performance
 - Requires expert opinion to determine what is good enough
 - Model = experiment
 - $A=0$



Probabilistic Validation - Error Metric

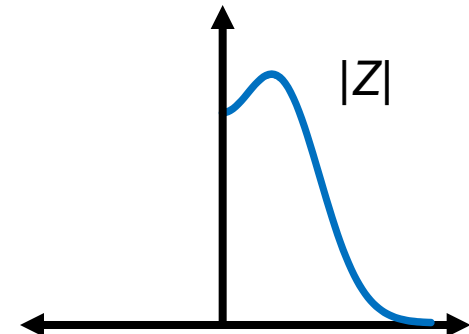
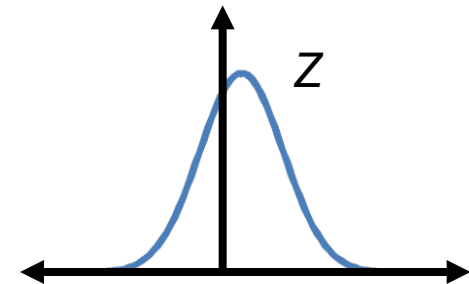
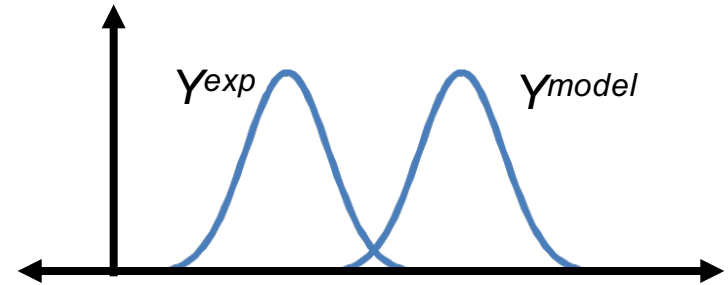
- Absolute error between a model prediction and an experimental response quantity
 - Model prediction and experimental measurement are uncertain
 - Normalized by the experimental mean value (to simplify solution)

$$Z = \frac{Y^{mod} - Y^{exp}}{E[Y^{exp}]}$$

$p = P(|Z| \leq z)$ Probability that the error will not be exceeded

- Validation Requirement

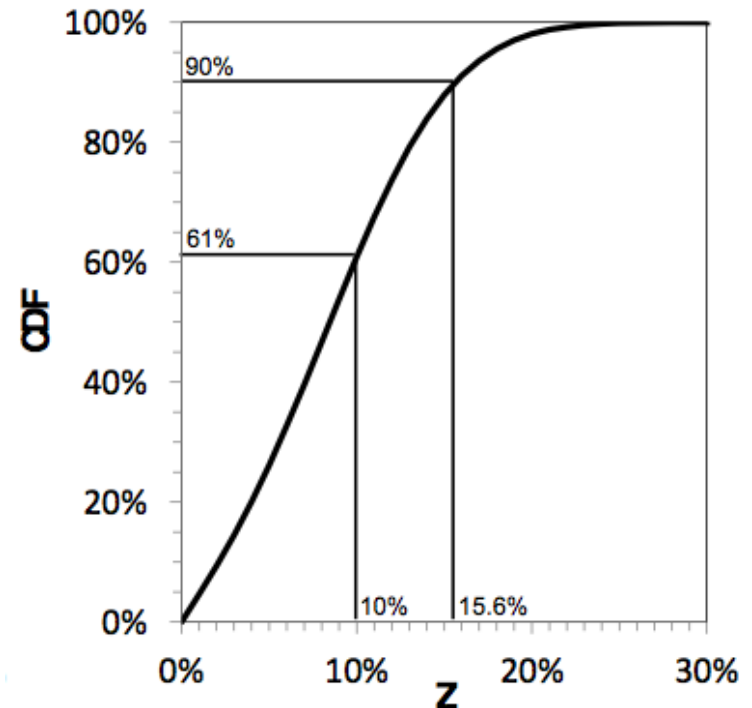
$$p < p_r, \text{ or } z < z_r$$



Thacker, B.H. and T.L. Paez, "A Simple Probabilistic Validation Metric for the Comparison of Uncertain Model and Test Results," AIAA SciTech, 16th AIAA Non-Deterministic Approaches Conference, National Harbor, Maryland, 13-17 January 2014.

Probabilistic Validation - Error Metric Interpretation

- CDF (integration of PDF) of Z
 - X-axis is error - Z
 - Y-axis is probability level - p
- 90% probability that the error will not be greater than 15.6%
- 61% probability that the error will not be greater than 10%
- The error between the model and the experiment is fully defined
- The benchmark level of error is the error of the experiment compared to itself at a 90% probability level

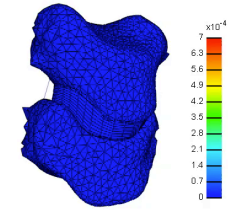


Z = error between model and experiment

Hierarchical Model V&V

Component Validation

With_nucleus.xplt
Z - displacement
Time = 0



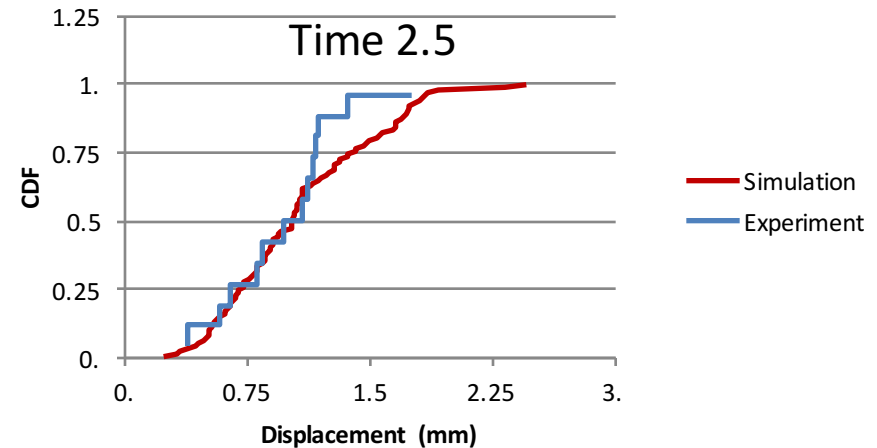
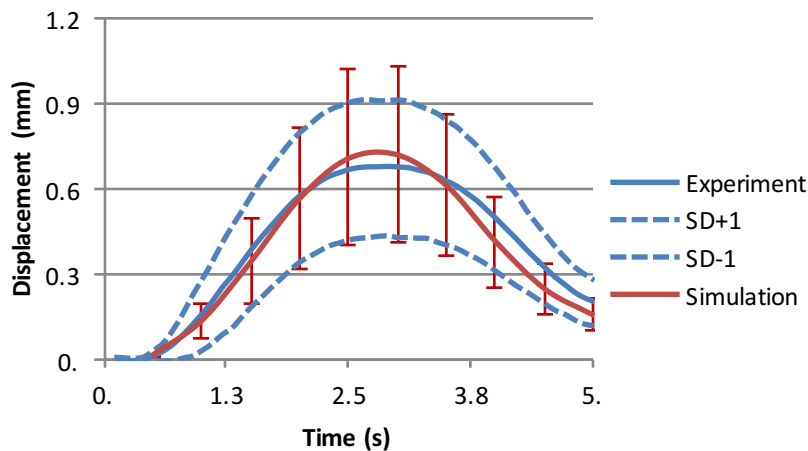
Experimental setup

- Isolated vertebra-disc-vertebra specimens
- Loaded with 100 N tension in sine function

Simulation

- Perform probabilistic analysis incorporating variability of material properties and viscoelastic properties
- Calculate area metric and z-metric

Time	Benchmark Error	Sim Validation	Area Metric
2.5	84%	82%	0.137
5	89%	88%	0.180



Example Component Level Validation

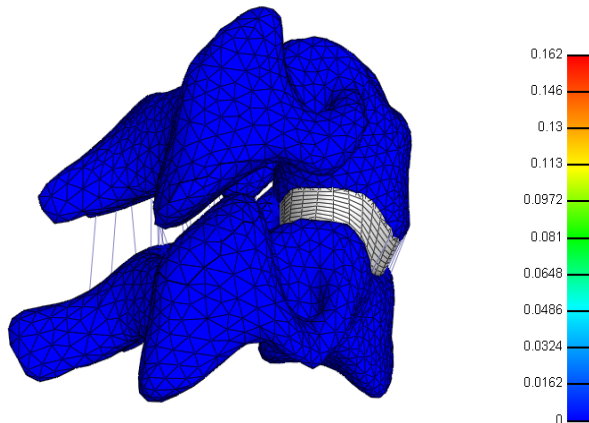
Experimental set-up

- Pure moment loading of complete motion segment to 2 N-m
- Rotation of superior vertebrae recorded

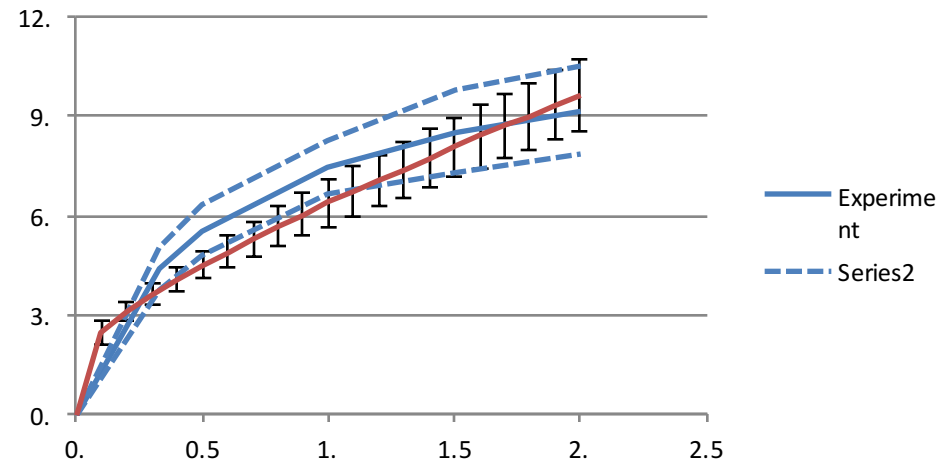
Simulation

- Perform probabilistic analysis incorporating variability of material properties (Disc and all ligaments)
- Calculate area metric and z-metric

flexion.xpt
X - rigid angular position
Time = 0



Load	Benchmark Error	Sim Validation	Area Metric
2 N-m	34%	93%	0.103



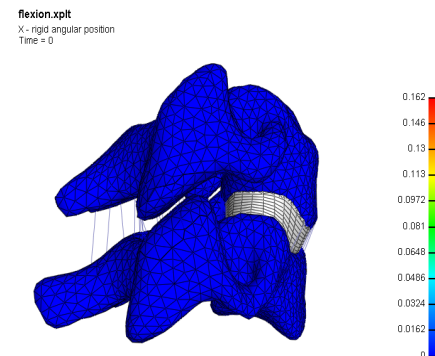
Example Component Level Validation

Experimental set-up

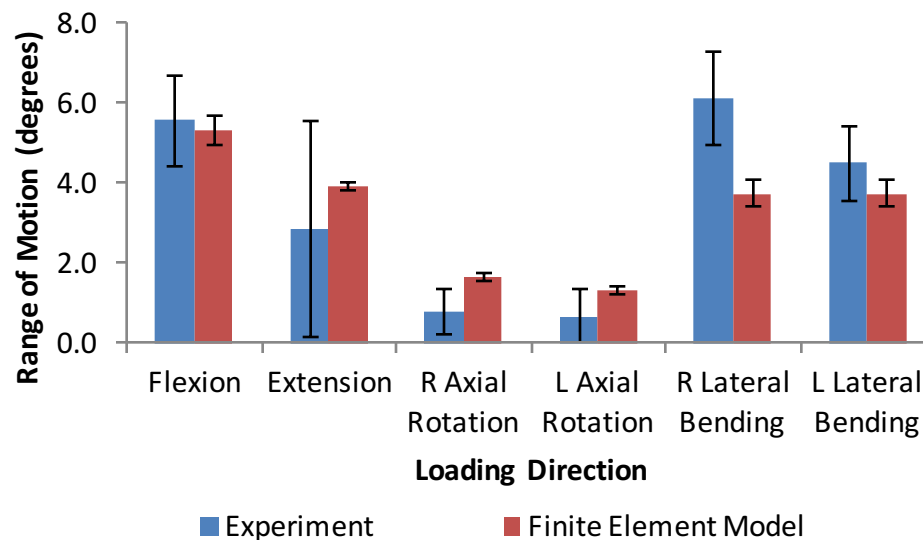
- Pure moment loading of complete motion segment to 6 N-m
- Rotation of superior vertebrae recorded

Simulation

- Perform probabilistic analysis incorporating variability of material properties (Disc and all ligaments)
- Calculate area metric and probabilistic error metric

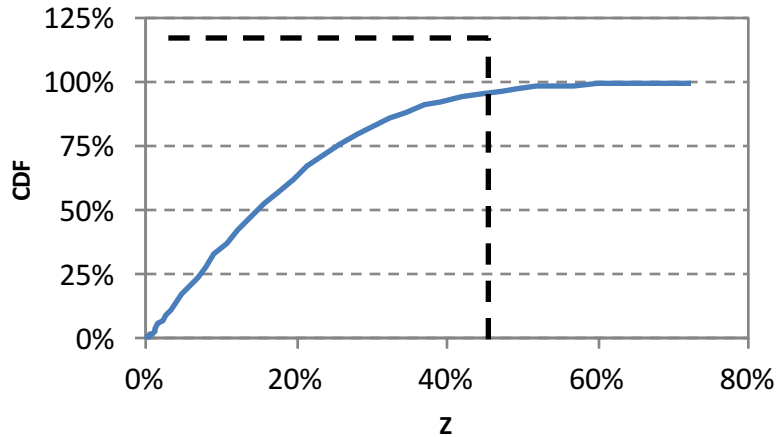


Qualitative
Validation

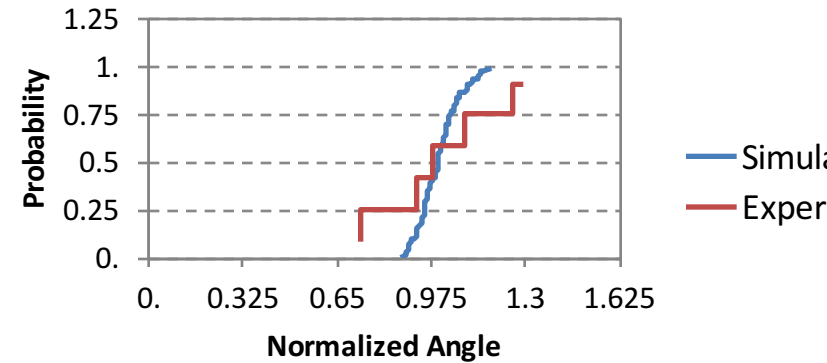


Example Component Level Validation

Probabilistic error metric



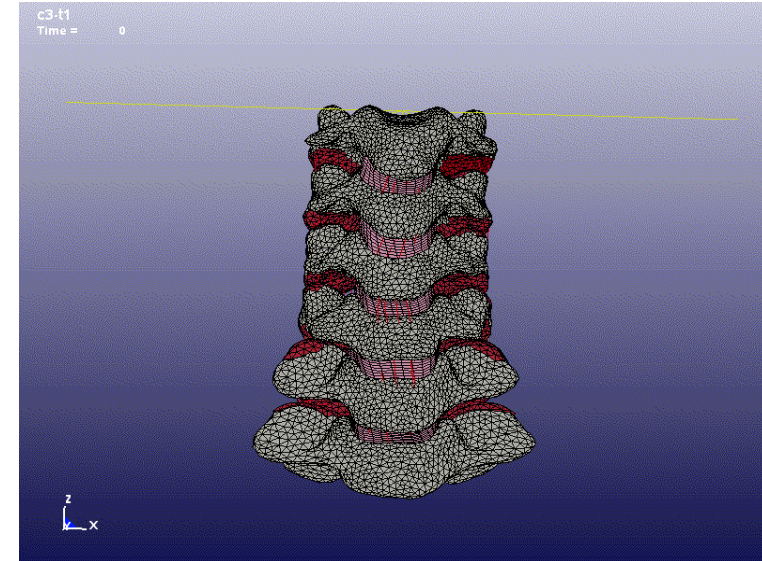
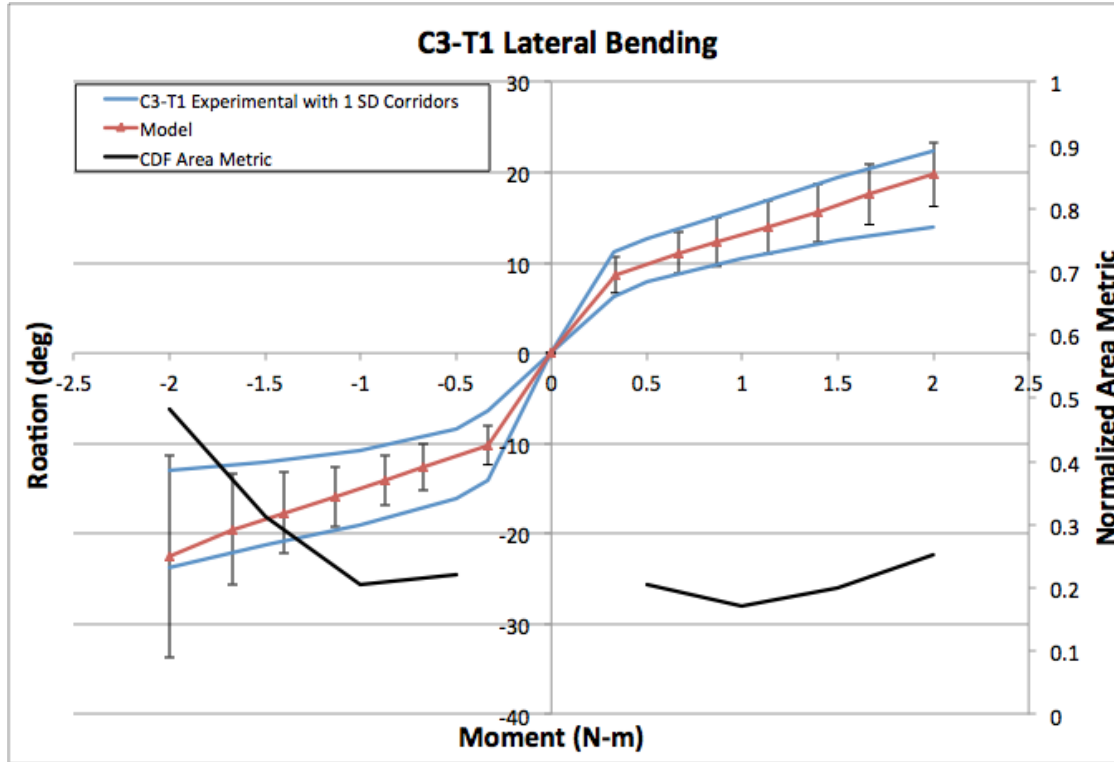
Area metric



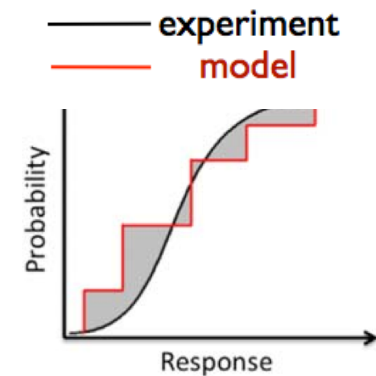
Motion	$e_{\text{benchmark}}$	FEM $p = P[Z \leq e_{\text{benchmark}}]$
Flexion	48%	98%
Extension	222%	97%
R Axial Rotation	172%	81%
L Axial Rotation	260%	91%
R Lateral Bending	45%	61%
L Lateral Bending	48%	77%

Motion	Area Metric
Flexion	0.11
Extension	0.60
R Axial Rotation	0.61
L Axial Rotation	0.67
R Lateral Bending	0.54
L Lateral Bending	0.17

Example Subsystem Level Validation



similar results for axial rotation, flexion, and extension

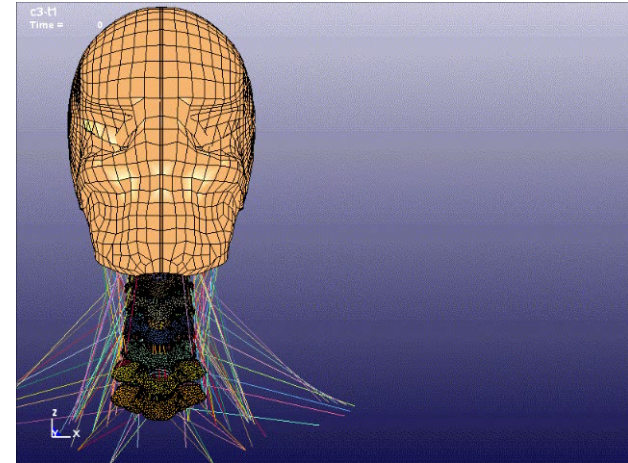


Model Verification and Validation

System Level V&V: Dynamic Lateral Impact

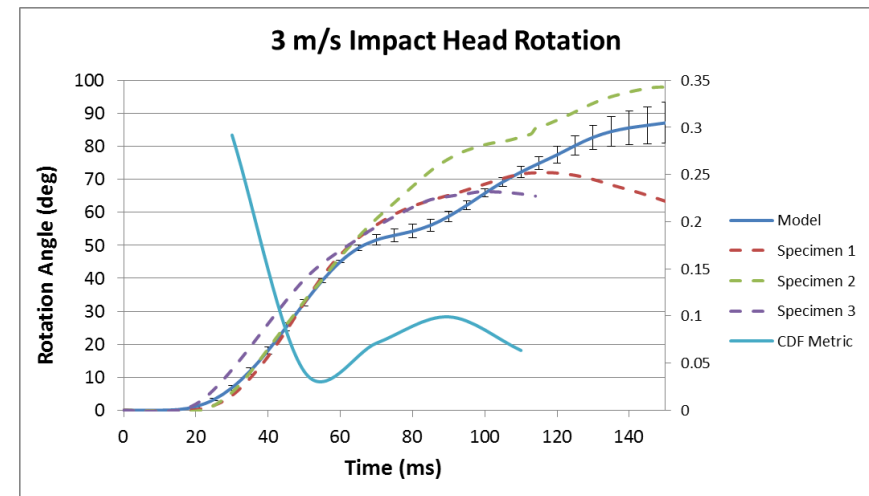
Experimental Setup

- Three PMHS specimens potted at T1 and mounted to lateral impact loading sled
- Sled was accelerated using a pendulum impactor at 1, 2, and 3 m/s impact velocities
- Head and vertebral kinematics were recorded using a Vicon motion capture system
- Accelerometer used to record sled accelerations



Simulation

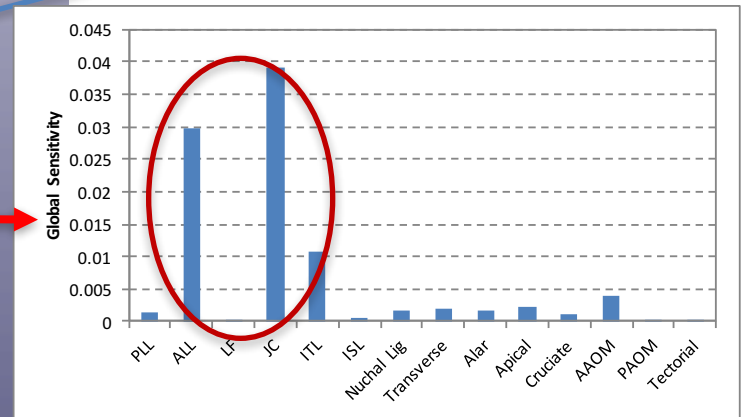
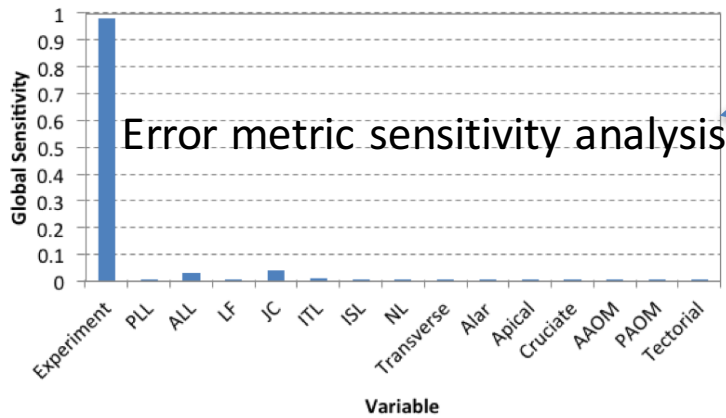
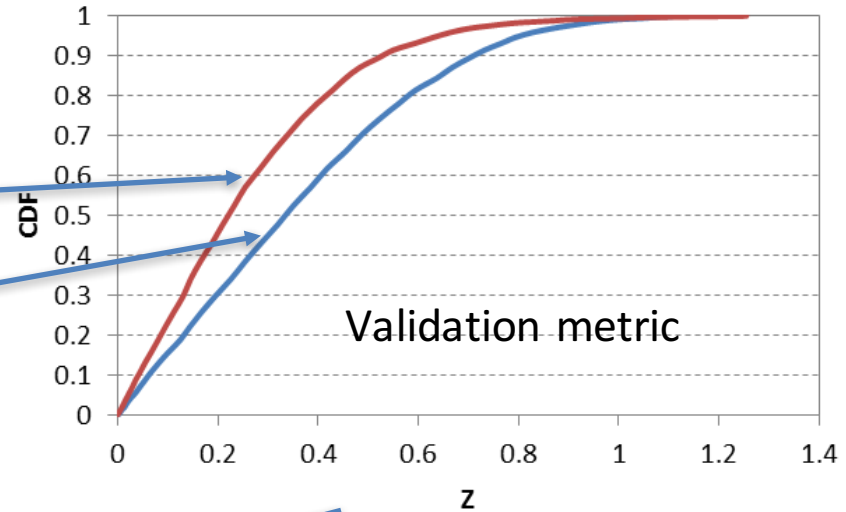
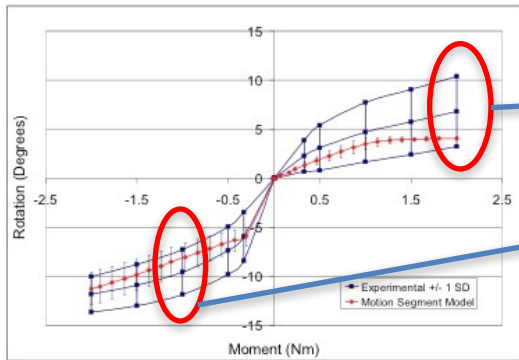
- Full Cervical spine + head
- Accelerations applied to T1
- Probabilistic analysis performed
- Head kinematics validated using area metric



Model Validation Example

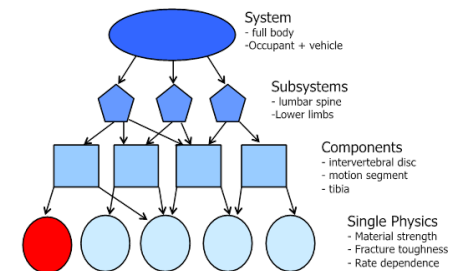
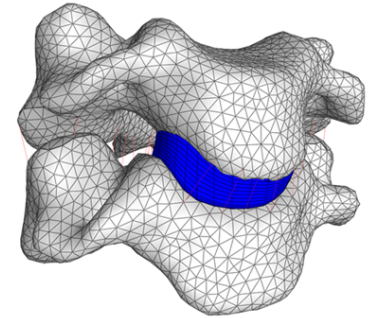
Sensitivity of Error to Model and Experiment Uncertainties

model vs. experiment

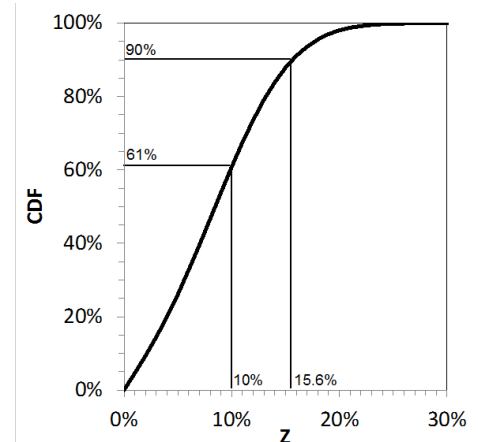


Summary

- Modeling tools allows the development of complex, high fidelity models
 - Model fidelity \neq model validity
- Hierarchical approach (ASME V&V-10 Committee)
 - Breaks the problem into smaller parts
 - Validation process employed for every element in the hierarchy (ideally)
 - Allows the model to be challenged (and proven) step by step
 - Dramatically increases likelihood of right answer for the right reason
- Modeling and experiment teams need to work together
 - Experiments should be designed for model validation
- Account for uncertainty in both model and experiment
- Validation metric is the measure of the mismatch between model and experiment – Quantitative
- Sensitivity analysis can provide some insight into source of mismatch



Approach: Bottoms up guided by sensitivity analysis of (un-validated) full system model





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Thank You

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