

IS THERE A PLACE FOR GENERIC MODELING UNCERTAINTY IN ENGINEERING DESIGN?

by

**Timothy K. Hasselman
ACTA Inc., Torrance, CA**

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Questions Posed

- ☐ How should we view model validation *in the context of engineering design*?
- ☐ How should we assess the *predictive* capability of a model in the absence of experimental data?
- ☐ How can we improve our confidence in predictive models for engineering design?
- ☐ What are the *open* critical research topics in model validation?
- ☐ Will we ever be able to design, develop and certify engineering systems in a pure virtual (computational) environment?

Requirements for Validation of Predictive Models Used in Engineering Design

- ❑ **Model validation for engineering design will require a process compatible with and accepted by the design establishment of an organization, where:**
- ❑ **The intended use of a model is clearly specified**
- ❑ **Comparative metrics are defined**
 - Generic global metrics for uncertainty quantification – e.g. normalized modal metrics, principal components, normalized Fourier coefficients, etc.
 - Specific local metrics for validation assessment – e.g. acceleration, velocity, displacement at points of interest, time to a critical event, etc.
- ❑ **Validation assessment is performed to evaluate the comparative metrics.**
- ❑ **Acceptance criteria are defined, by which the adequacy of a model validation assessment *for the intended use of the model* is judged.**

Some Options for Assessing Predictive Capability in the Absence of Experimental Data

☐ Use prior experience alone, based on analysis-test comparisons

- Accepted uses: predictive accuracy of modal frequencies for control system design
- Suggested applications: predictive accuracy of complete analytical models for
 - Truss-type space structures
 - Launch vehicle payloads
 - Automobile crashworthiness simulation

☐ Use a Bayesian approach with prior experience

- Update the covariance of model response metrics based on estimated parameter uncertainties, with a generic covariance matrix of analysis-test differences based on prior analysis and test experience.
- Update both the mean and covariance of model response metrics based on estimated parameters and their uncertainties, with the estimated mean and covariance of analysis-test differences from prior analysis and test experience (requires more data in a narrower generic category).

☐ Use a Bayesian approach with prior experience based on “lessons learned”

- Same as previous approach except that estimated mean and covariance from the generic database are based on analytical models that have been updated to remove identified modeling errors.
- These modeling errors must qualify as “lessons learned” that are reflected in subsequent modeling efforts, including the current one.

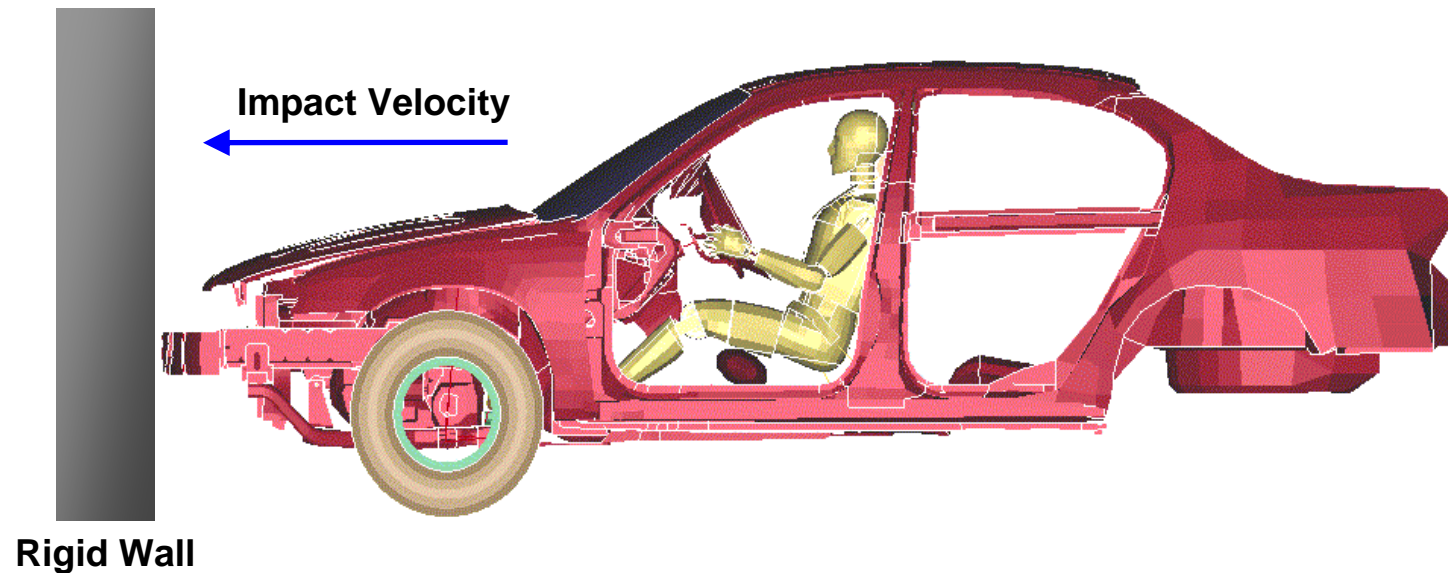
Example: Crashworthiness Tests and Simulations (Courtesy GM)

❑ Crashworthiness Tests

- Midsize test vehicle
- Acceleration and force
- Measurements on vehicle, barrier, dummy
- Straight frontal, left/right angle barrier
- SDM, RadR, RadL, RadC

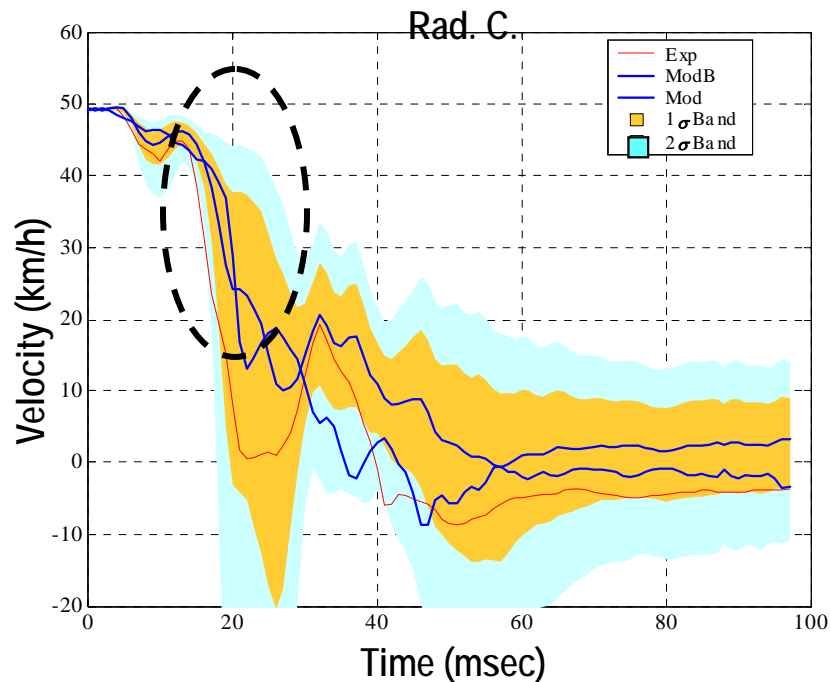
❑ Numerical Simulations

- LS-DYNA
- Velocity time-history
- At common locations

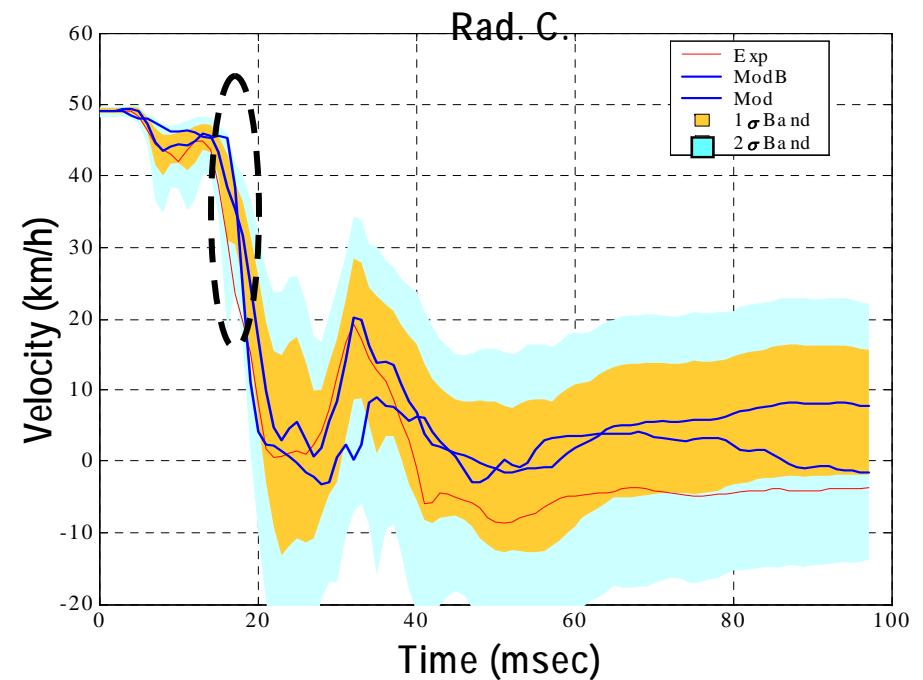


Reduction of Modeling Uncertainty

Prior to Model Improvement



After Model Improvement



Courtesy GM and 2005 IMAC Paper No. 278

Improving Confidence in Predictive Models for Engineering Design

- ❑ Accumulate data to a generic historical database of analysis-test experience as predictive engineering design models are eventually validated by comparison with test data, i.e. build a “living” database.
- ❑ Take advantage of lessons learned in building future analytical models and compiling the historical database.
- ❑ As more data become available, attempt to narrow generic categories, thereby reducing generic modeling uncertainty.
- ❑ Update the *process* used for validating predictive models used in engineering design by periodically evaluating the performance of the process against project goals.

Critical Research Topics

❑ With regard to the use of generic modeling uncertainty in engineering design to assess the predictive accuracy of pretest models, critical research topics might include:

- Development of guidelines for the classification of generic categories, e.g. w.r.t. crashworthiness simulation of automobiles, one might consider size (compact, mid-size, full-size); materials (steel, aluminum, composites), model fidelity, test fidelity, etc.
- Trade-offs between broader categories and more data sets and narrower categories with fewer data, in terms of statistical significance and sampling-based confidence levels.

❑ In a broader context, one might consider in the investigation of alternative approaches:

- What resources are available within an organization to support various approaches, e.g. availability of required engineering data (both analysis and test data)?
- What approach best fits the design engineering culture of an organization?
- How long would it take to realize the benefits of a given approach?
- Does the organization have the financial and management resources to make the necessary commitment to an engineering design validation program?

Looking to the Future: Can we get there from here?

Can we ever design, develop and certify engineered systems in a virtual computational environment? Some suggestions:

- ☐ **Select an approach; outline a program with a pilot project.**
- ☐ **Demonstrate added value in the pilot project - possibly two design cycles, 2 – 5 years**
- ☐ **Estimate the short-term and long-term benefits of the program.**
- ☐ **Estimate the cost of a long-term commitment to the program.**
- ☐ **Establish goals and cost/benefit metrics so that progress toward those goals can be objectively monitored.**
- ☐ **Schedule periodic program reviews with built-in opportunities for mid-course adjustments.**