

Validation Issues Associated with Unsteady Flows

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Outline

- **Traditional vs. validation experiments**
 - Goals of model validation
 - Trade-off in model validation
 - Design of validation experiments
- **Structure of computational simulation**
 - Non-deterministic simulation
 - Validation versus calibration
- **Areas in need of improvement**
 - Experimental activities
 - Computational activities
- **Conclusion and recommendations**

Traditional Experiments vs. Validation Experiments

Goals of three types of traditional experiments:

1. Improve the fundamental understanding of the physics:

- Ex: fluid dynamic turbulence experiment; experiment for understanding evaporation and condensation in multi-phase flows

2. Improve the mathematical models of some physical phenomena:

- Ex: experiment to determine reaction rate parameters in reacting flow; experiment for calibrating parameters in two-equation turbulence models

3. Assess subsystem or complete system performance:

- Ex: performance of a new combustor design in a gas turbine engine; tests of a new multi-element flap design for a wing

- **Model validation experiment**

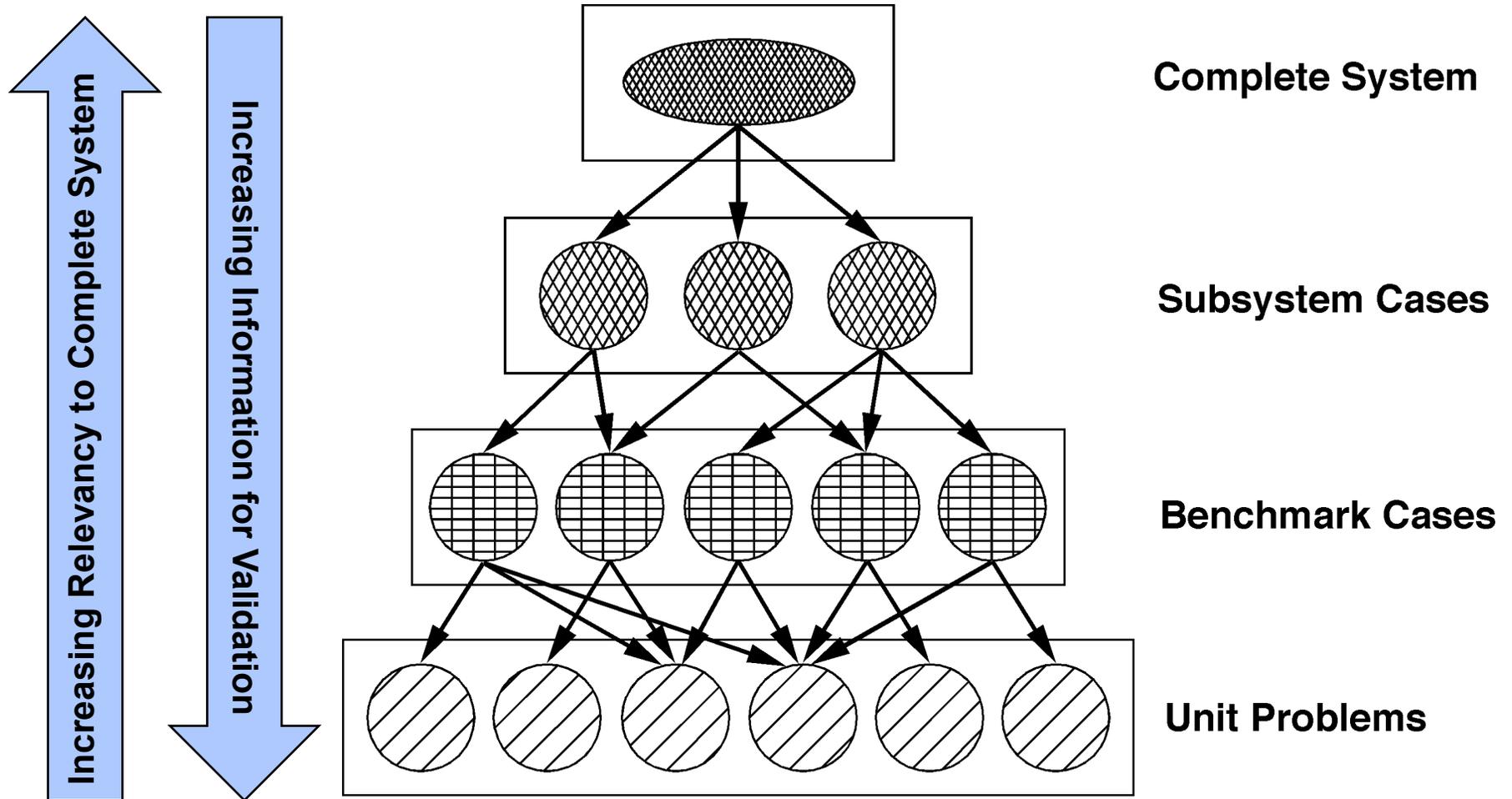
- **An experiment that is designed and executed to quantitatively estimate a mathematical model's ability to simulate a physical system or process.**

- **The computational model developer or code user is the customer.**

Goals of Model Validation

- **Tactical goals of validation:**
 - Quantification of the effects of mathematical modeling assumptions and approximations by comparison of simulation results with experimental measurements
 - Identification of the effect of physics modeling weaknesses, i.e., quantification of model form uncertainty
 - Model form uncertainty is distinct from model parameter uncertainty
- **Strategic goals of validation:**
 - Improve the separation of model form uncertainty from all other forms of uncertainty, particularly model parameter uncertainty
 - Improve physics modeling to improve predictive capability
- **What are the **crucial elements** of model validation?**
 - Experimental measurement of the important input data for the model
 - Validation metrics: mathematical operators that quantify the difference between simulation and experimental results

Trade-Off in Model Validation Experiments

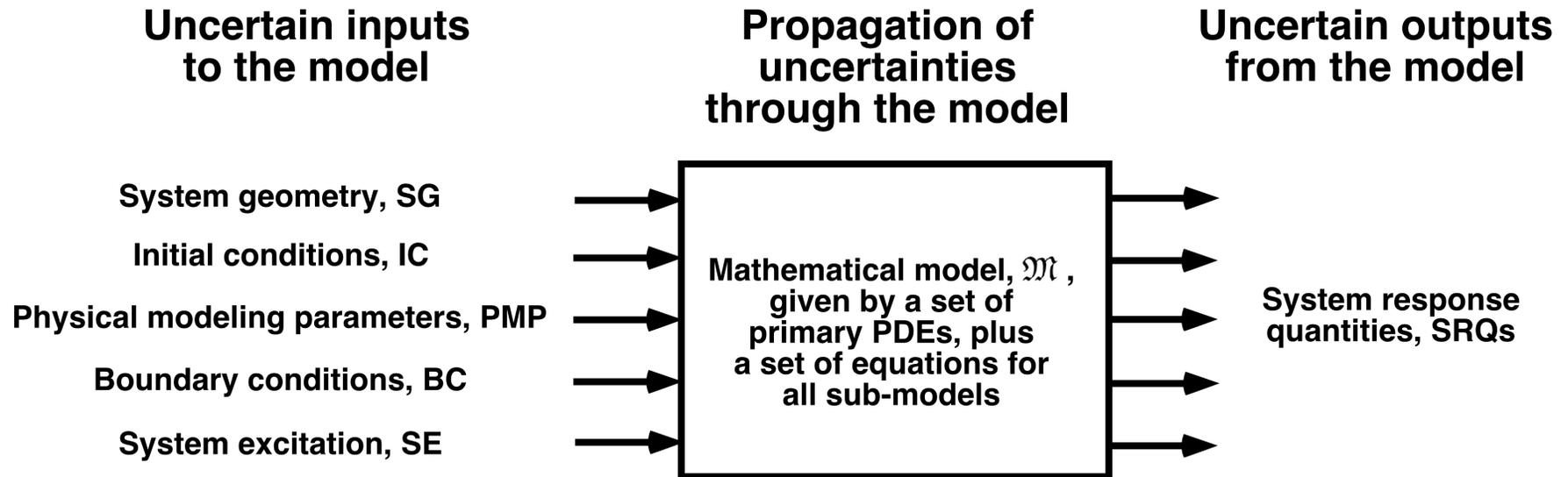


(Ref: AIAA Guide, 1998)

Design of Validation Experiments

- **Key elements in the design of validation experiments:**
 - Modelers and experimentalists **jointly** design the experiment
 - Experimentalist should measure all important model input data
 - Achievement of a blind prediction is the most effective approach
 - Experimentalist should estimate uncertainty on **both** input data and system response quantities measured
- **Five categories of input data:**
 - System geometry
 - Initial conditions
 - System physical parameters
 - Boundary conditions
 - System excitation
- **What makes validation particularly difficult?**
 - Connecting modelers and experimentalist, on the same time frame
 - Separation of input uncertainty and model form uncertainty

Structure of Computational Simulation



- **Examples of uncertainty in input data for validation experiments:**
 - Geometry; detailed measurement of the actual as-tested geometry
 - Initial conditions; flow field at the beginning of engine-unstart
 - Physical parameters; non-equilibrium chemistry in a high enthalpy facility
 - Boundary conditions; spatial and temporal characterization of inflow conditions in a wind tunnel
 - System excitation; acoustic excitation of a turbulent boundary layer

Where Do We Stand in Model Validation?

- **Common approach to validation is actually **model calibration**:**
 - Input data and parameters in the model (either scalars or probability distributions) are calibrated to improve agreement with experimental data
 - When model validation (i.e., model accuracy assessment) is mixed with model calibration, it underestimates model form uncertainty by adjusting input data uncertainty
- **To improve confidence in our simulations, validation should:**
 - Improve the separation of calibration and validation activities
 - Emphasize the assessment of simulation accuracy by using blind-predictions of experimental data
 - Improve cooperation and teamwork between experimentalists and computational analysts so as to conduct improved validation experiments

Areas in Need of Improvement: Experimental Activities

- **Improved spatial and temporal characterization of inflow boundary conditions for CFD simulation**
- **Improved spatial and temporal measurement of flow in an empty test section of a wind tunnel, including the tunnel walls**
- **Improved measurement and documentation of time-dependent quantities, especially in high-enthalpy shock tunnels**
- **Comparison of measurements using different experimental techniques, e.g., PIV versus LDV**
- **Improved experimental uncertainty estimation by way of:**
 - **Comparison of run-to-run and facility-to-facility variability**
 - **Comparison of runs with the test article at different locations in the test section**
- **Improved assessment of experiments regarding completeness of information for model validation**

Areas in Need of Improvement: Computational Activities

- Use of simulation to guide the design and execution of validation experiments, e.g., use of sensitivity analyses
- Improved code verification testing and documentation
- Improved quantification and documentation of numerical error due to temporal and spatial discretization
- Willingness to compute the flow in the entire test section
- Construction of improved validation metrics for unsteady flows
- Improve cooperation and teaming with experimentalists

This is the responsibility of management and funding sources

Conclusions and Recommendations

- “Quality” of a validation experiment should be ranked on the following characteristics (in priority order):
 - Measurement and documentation of important input data needed for the simulation
 - Estimation of experimental uncertainty on both input and output data
 - Assessment of model accuracy by way of a blind prediction
 - Separation of model calibration and model validation
- Validation is focused on simulation of the flow field **inside the wind tunnel or in flight**
- The burden of improving the quality of validation experiments falls primarily on experimentalists
- Sponsors should recognize the need to fund new validation experiments

This is the path to critically assessing our simulations

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