

An Overview of ASME V&V 20: Standard for Verification and Validation in Computational Fluid Dynamics and Heat Transfer

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OUTLINE

- **Origin of the approach; background**
- **ASME V&V 20 -- Overview**

Origin of this V&V Approach

- US Office of Naval Research Program 1996-2000 – Produce a “documented solution”
 - Could unsteady RANS research codes be implemented with confidence in the design of the next generation of naval vessels?
 - Experiments on models in 3 towing tanks in U.S. (David Taylor, IIHR) and Italy (INSEAN)
 - Two RANS codes used by (a) code developers and (b) other groups
 - Classified program
- A quantitative V&V approach was proposed based on error and uncertainty concepts in experimental uncertainty analysis (ISO GUM, 1993, international standard).
 - Hugh Coleman (UAHuntsville) and Fred Stern (Iowa) published initial version in ASME Journal of Fluids Engineering, Dec 1997.

V&V 20 Development

ASME Performance Test Codes Committee PTC 61:

V&V 20: Standard for Verification and Validation in Computational Fluid Dynamics and Heat Transfer

Approach is based on experimental uncertainty analysis concepts of error and uncertainty. Committee formed in 2004; Draft document completed; peer review comments received early June 2008; publication probable in late 2008.

Hugh Coleman, UAHuntsville, Chair

Chris Freitas, SwRI, Vice-Chair

Glenn Steele, Miss. State Univ.

Patrick Roache, Consultant

Urmila Ghia, Univ. Cincinnati

Ben Blackwell, Consultant (retired Sandia-ABQ)

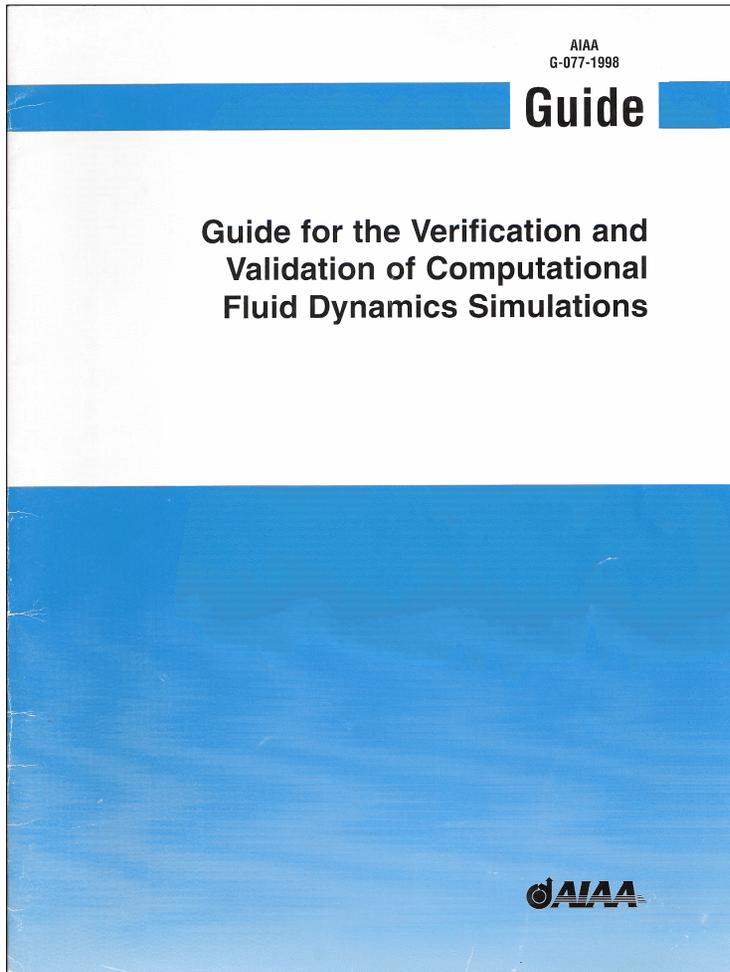
Kevin Dowding (Sandia-ABQ)

Richard Hills, New Mexico State Univ.

Roger Logan, Lawrence Livermore Nat'l Lab.

How Does V&V 20 Fit With Previously-Published V&V Guides ?

- AIAA CFD Standards Committee: AIAA Guide G-077-1998 “Guide for the Verification and Validation of Computational Fluid Dynamics Simulations”
- ASME PTC 60: V&V 10 (2006) “Guide for Verification and Validation in Computational Solid Mechanics”
- ASME PTC 61: V&V 20 (2008) “Standard for Verification and Validation in Computational Fluid Dynamics and Heat Transfer”



•AIAA CFD Standards Committee: AIAA Guide G-077-1998 “Guide for the Verification and Validation of Computational Fluid Dynamics Simulations”

–Error: A recognizable deficiency in any phase or activity of modeling and simulation that is not due to lack of knowledge.

–Uncertainty: A potential deficiency in any phase or activity of the modeling process that is due to lack of knowledge.

ASME V&V 10-2006

Guide for Verification and Validation in Computational Solid Mechanics

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•ASME PTC 60: V&V 10 (2006) “Guide for Verification and Validation in Computational Solid Mechanics”

–Error: A recognizable deficiency in any phase or activity of modeling or experimentation that is not due to lack of knowledge.

–Uncertainty: A potential deficiency in any phase or activity of the modeling or experimentation process that is due to inherent variability or lack of knowledge.

ASME V&V 20-2008

**Standard for
Verification and Validation
in
Computational Fluid
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and
Heat Transfer**

(to appear, 2008)

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The American Society of
Mechanical Engineers

- The objective of V&V 20:
the specification of an approach that quantifies the degree of accuracy inferred from the comparison of solution and data for a specified variable at a specified validation point.

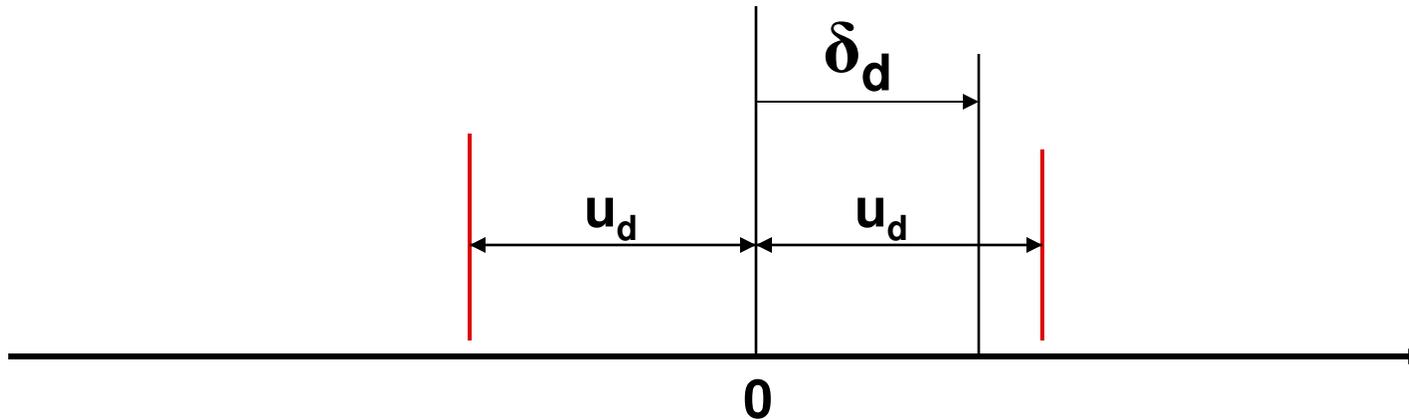
- The scope of V&V 20:
the quantification of the degree of accuracy for cases in which the conditions of the actual experiment are simulated.

“How good is the prediction? What is the modeling error?” --- at the validation point --- when the experiment itself is simulated.

Experimental Uncertainty Concepts: Error and Uncertainty

An **error** δ is a quantity with a sign and magnitude. A specific error δ_i is the difference (caused by error source i) between a quantity (measured or simulated) and its true value. (We assume there has been a correction made for any error whose sign and magnitude is known, so the errors that remain are of unknown sign and magnitude.)

An **uncertainty** u_i is an estimate of an interval $\pm u_i$ that should contain δ_i . (A standard uncertainty u is an estimate of the standard deviation of the parent distribution of δ : ISO GUM)

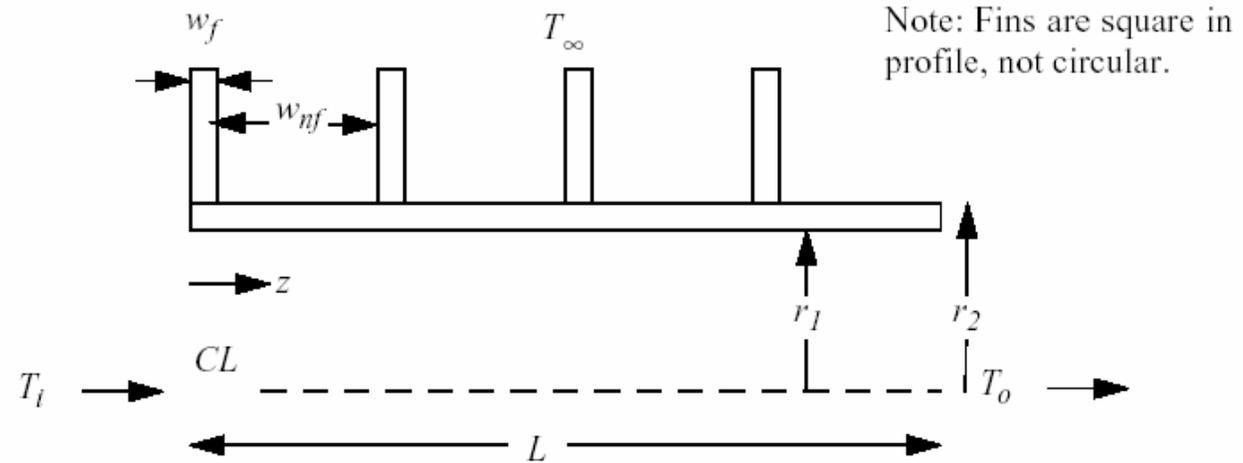


For example, for an (unknown) error δ_d in the data, u_d would be the standard uncertainty estimate.

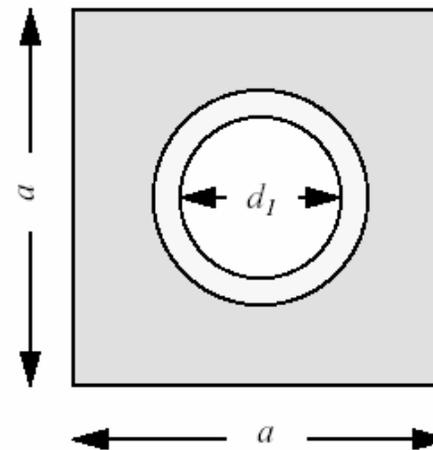
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Example for V&V 20 Nomenclature and Approach



a: cross section

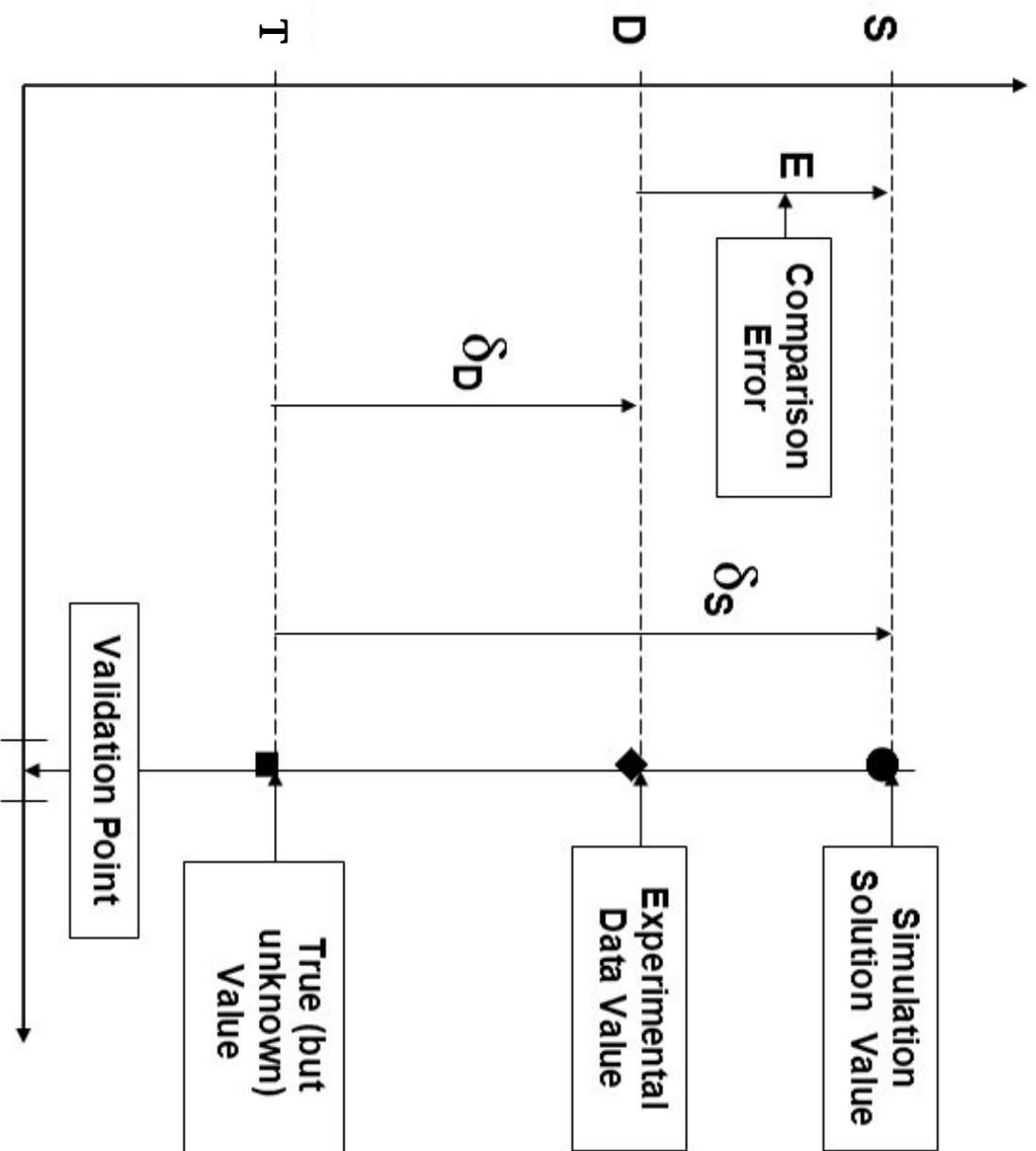


b: end view

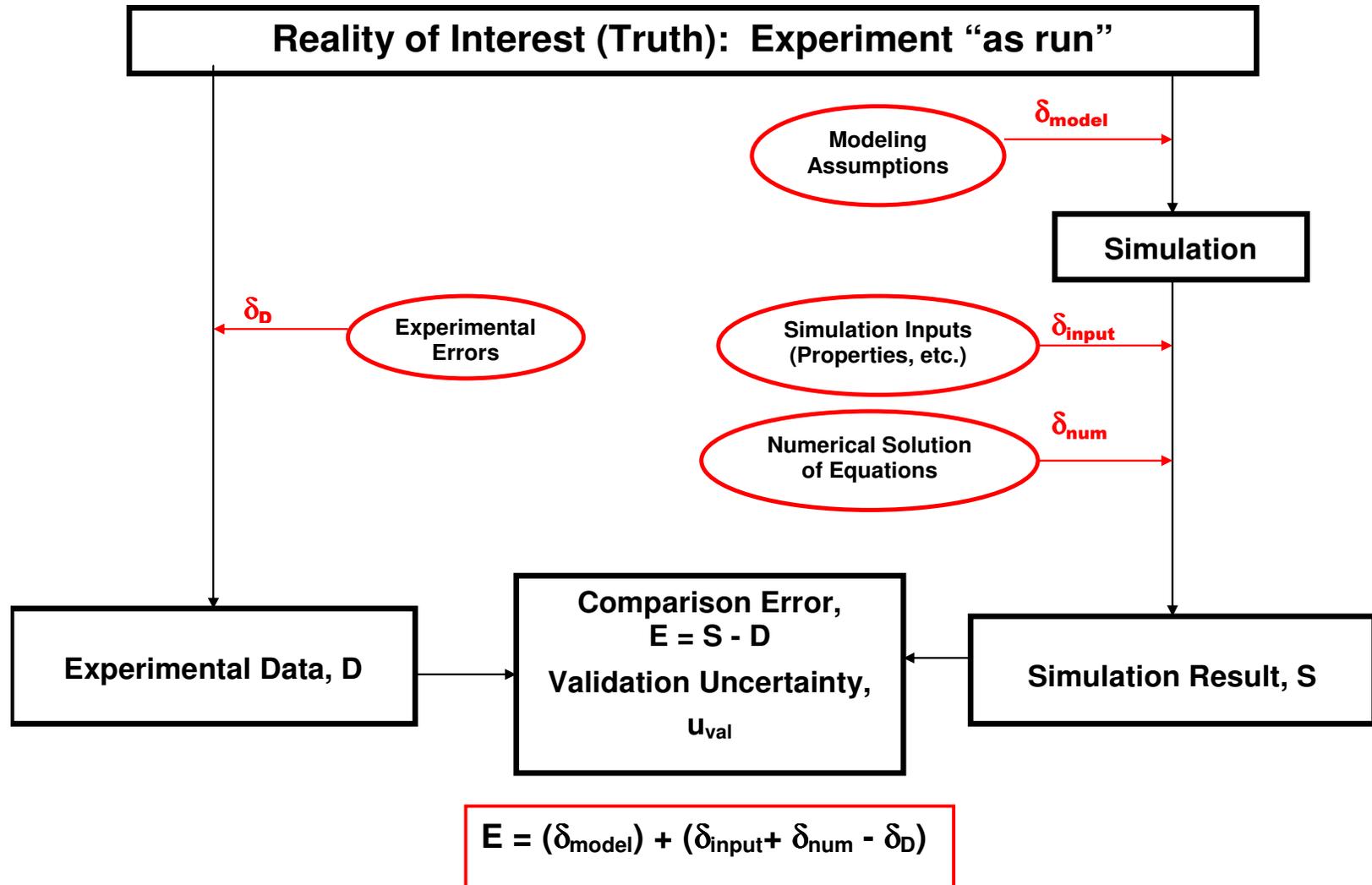
fin thickness = w_f

Validation variables of interest are T_o and $q = \rho Q(T_i - T_o)$

Outlet Bulk Fluid Temperature, T_o



A Validation Comparison



V&V Overview – Sources of Error Shown in Ovals

Strategy of the Approach

- Isolate the modeling error, having a value or uncertainty for everything else

$$E = \delta_{\text{model}} + (\delta_{\text{input}} + \delta_{\text{num}} - \delta_{\text{D}}) \longrightarrow \begin{array}{c} \overleftarrow{\hspace{10em} E \hspace{10em}} \\ \hline \end{array}$$

$$\delta_{\text{model}} = E - (\delta_{\text{input}} + \delta_{\text{num}} - \delta_{\text{D}})$$

- If $\pm u_{\text{val}}$ is an interval that includes $(\delta_{\text{input}} + \delta_{\text{num}} - \delta_{\text{D}}) \longrightarrow \begin{array}{c} \overleftarrow{\hspace{10em} \pm u_{\text{val}} \hspace{10em}} \\ \hline \end{array}$

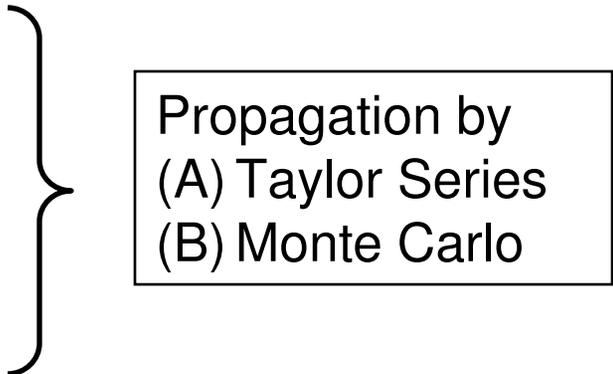
then δ_{model} lies within the interval

$$E \pm u_{\text{val}}$$



Uncertainty Estimates Necessary to Obtain the Validation Uncertainty u_{val}

$$u_{val} = \left(u_D^2 + u_{num}^2 + u_{input}^2 \right)^{1/2}$$

- Uncertainty in simulation result due to numerical solution of the equations, u_{num} (code and solution verification)
 - Uncertainty in experimental result, u_D
 - Uncertainty in simulation result due to uncertainties in code inputs, u_{input}
- 
- Propagation by
(A) Taylor Series
(B) Monte Carlo

Uncertainty Estimates Necessary to Obtain the Validation Uncertainty u_{val}

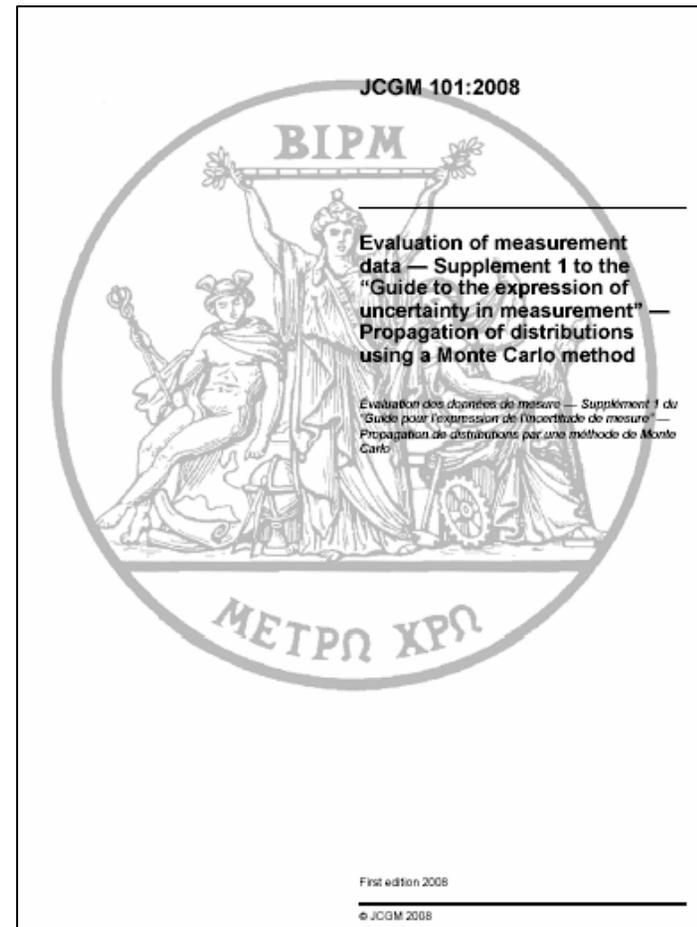
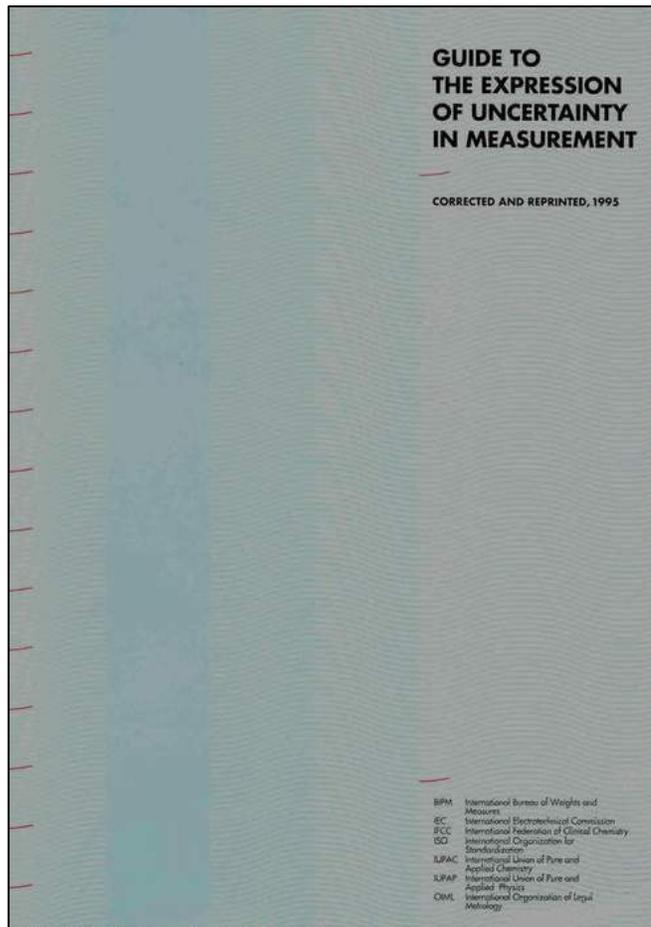
$$u_{val} = \left(u_D^2 + u_{num}^2 + u_{input}^2 \right)^{1/2}$$

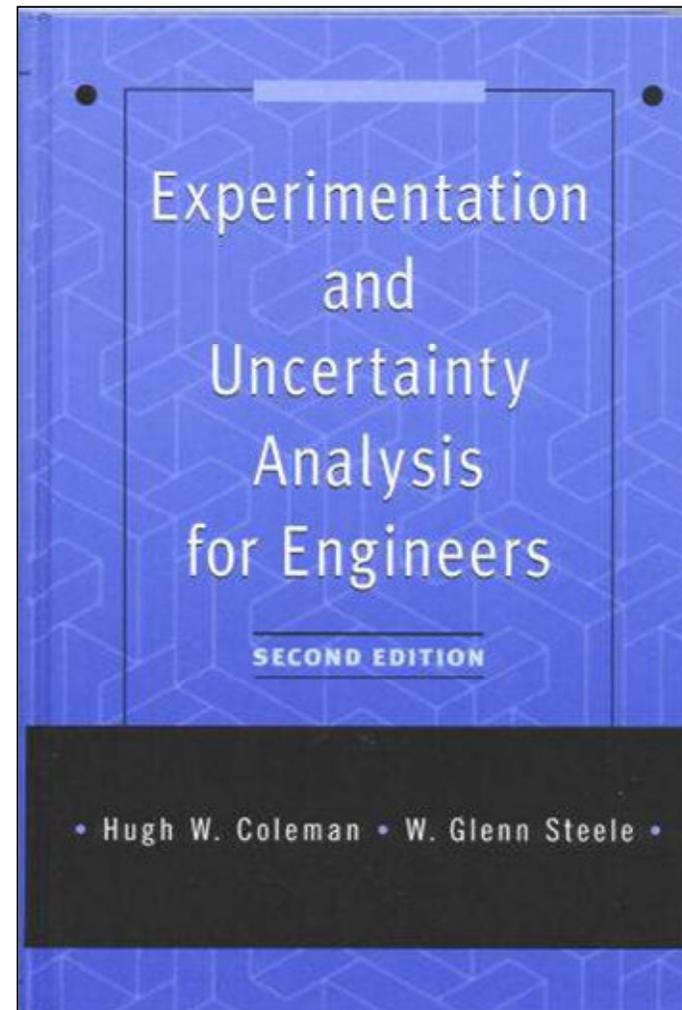
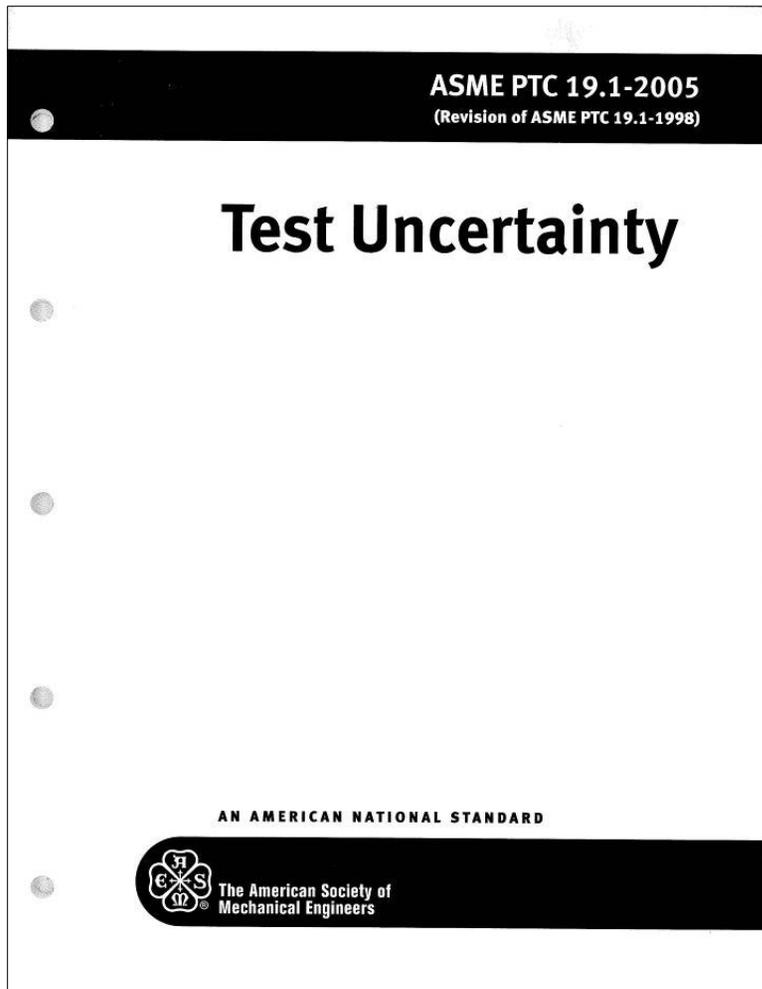
- **Code verification**: establishes that the code accurately solves the conceptual model incorporated in the code, i.e. that the code is free of mistakes for the simulations of interest. (MMS,)
- **Solution verification**: estimates the numerical accuracy of a particular calculation, i.e., u_{num} . (RE, GCI,)
- Eça, L., Hoekstra, M., and Roache, P. J. (2007), "Verification of Calculations: an Overview of the Second Lisbon Workshop," AIAA Paper 2007-4089.

Uncertainty Estimates Necessary to Obtain the Validation Uncertainty u_{val}

$$u_{val} = \left(u_D^2 + u_{num}^2 + u_{input}^2 \right)^{1/2}$$

- u_D can be estimated using experimental uncertainty analysis techniques





3rd Edition in 2009

Uncertainty Estimates Necessary to Obtain the Validation Uncertainty u_{val}

$$u_{val} = \left(u_D^2 + u_{num}^2 + u_{input}^2 \right)^{1/2}$$

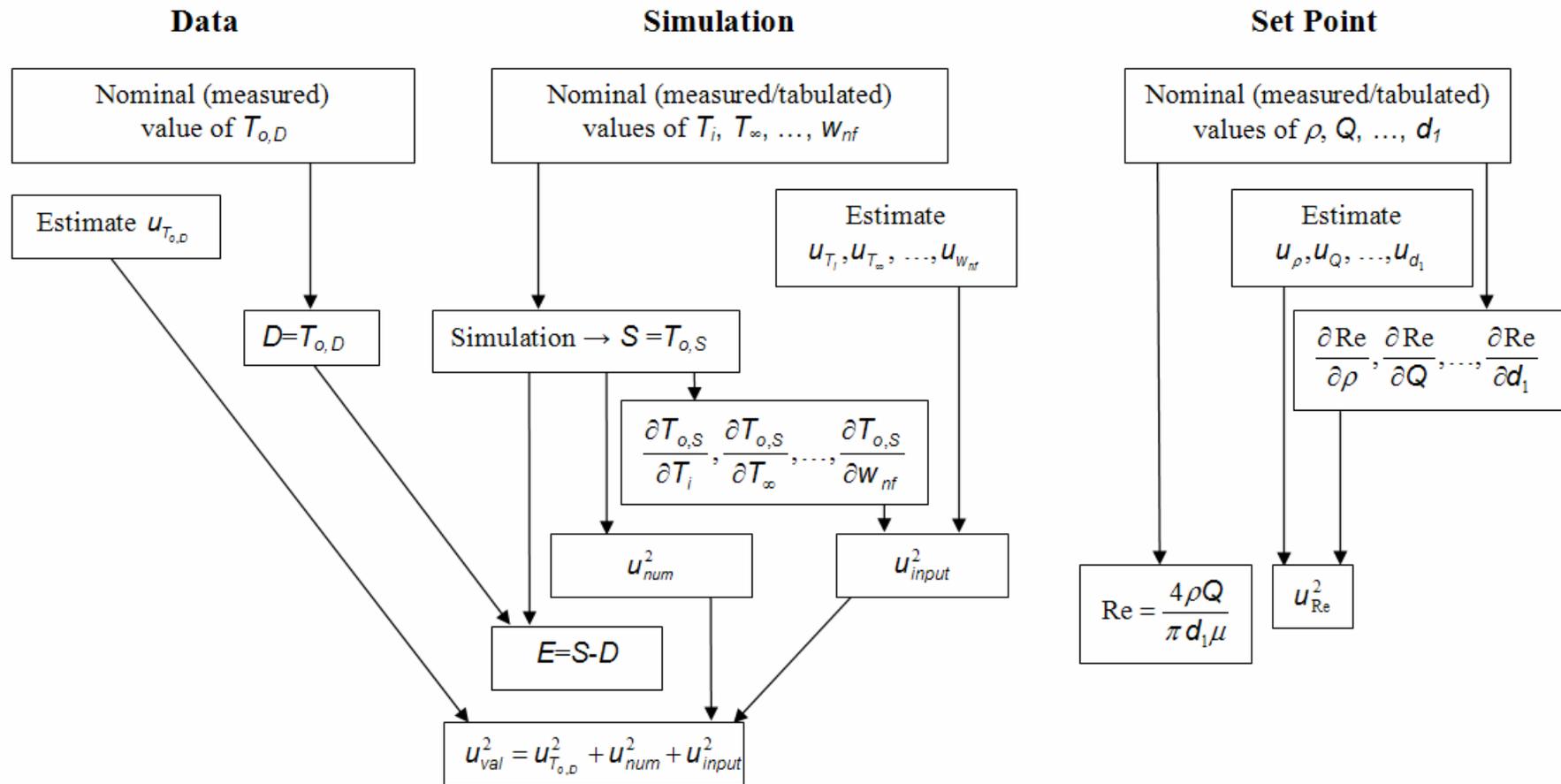
Taylor Series propagation approach to estimating u_{input}

$$u_{input}^2 = \sum_{i=1}^m \left(\frac{\partial S}{\partial X_i} \right)^2 (u_{X_i})^2$$

and the u_{X_i} are the uncertainties in the m simulation inputs X_i

(This expression for u_{input} is strictly true only when there are no shared variables in S and D . A more complex form is necessary if S and D contain shared variables, and is presented in detail in V&V 20)

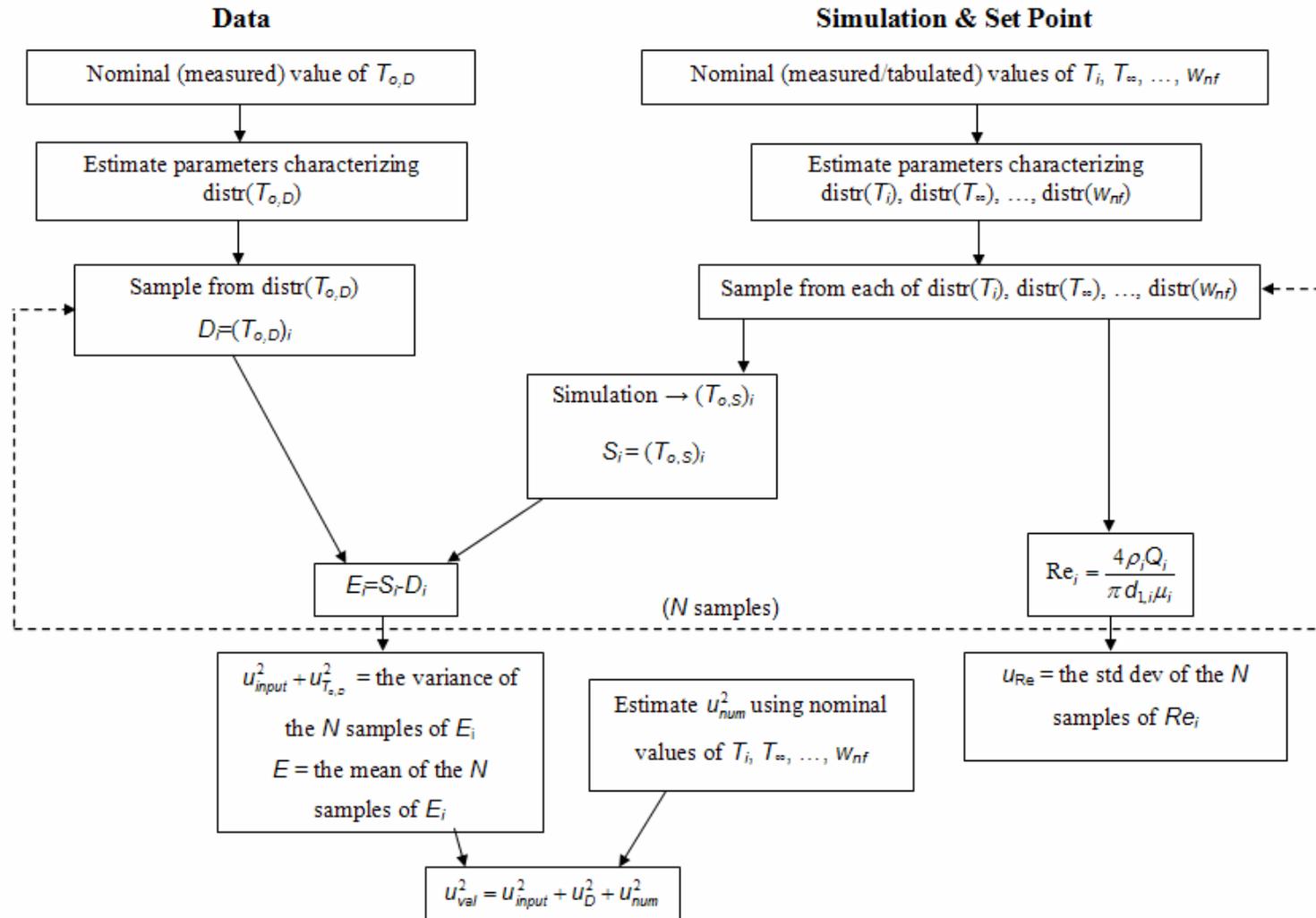
Taylor Series approach for estimating u_{val} when the validation variable T_o is directly-measured ($T_{o,D}$) and predicted with the simulation ($T_{o,S}$) as

$$T_{o,S} = T_{o,S}(T_i, T_\infty, Q, \rho, \mu, C_P, h_1, h_2, h_f, h_c, k_f, k_t, d_1, d_2, L, a, w_f, w_{nf})$$


Case 1

Monte Carlo approach for estimating u_{val} when the validation variable T_o is directly-measured ($T_{o,D}$) and predicted with the simulation ($T_{o,S}$) as

$$T_{o,S} = T_{o,S}(T_i, T_\infty, Q, \rho, \mu, C_P, h_1, h_2, h_f, h_c, k_f, k_t, d_1, d_2, L, a, w_f, w_{nf})$$



Case 1

Additional Cases Covered in V&V 20

- The experimental value D of the validation variable is determined from a data reduction equation

$$q_D = \rho Q C_P (T_{i,D} - T_{o,D})$$

and the simulation value predicted as

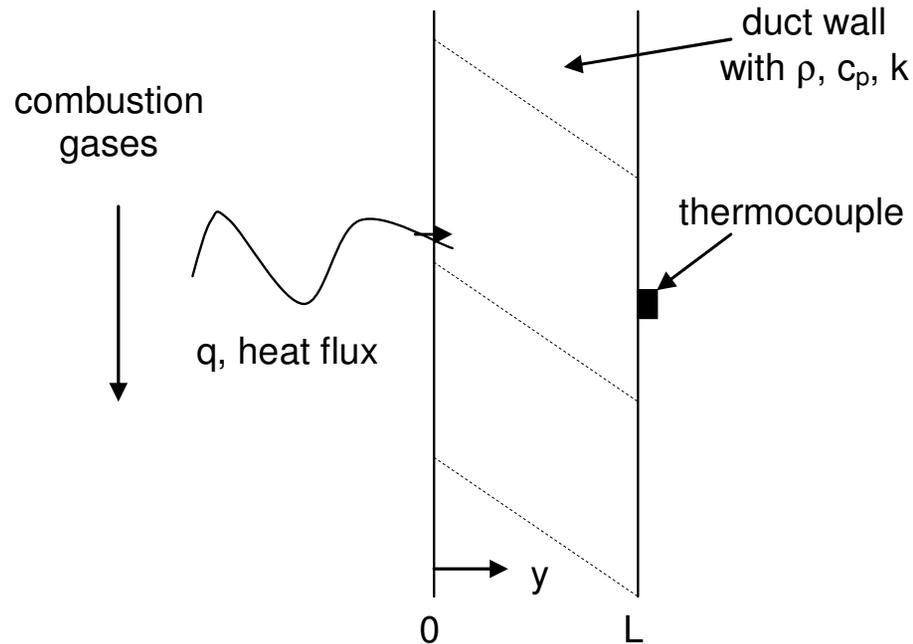
$$q_S = \rho Q C_P \left[T_{i,D} - T_{o,S} (T_i, T_\infty, Q, \rho, \mu, C_P, h_1, h_2, h_f, h_c, k_f, k_t, d_1, d_2, L, a, w_f, w_{nf}) \right]$$

V&V 20 Case 2: $T_{i,D}$ and $T_{o,D}$ share no error sources, so there are no correlated systematic errors

V&V 20 Case 3: $T_{i,D}$ and $T_{o,D}$ are measured with transducers calibrated against the same standard, so there are correlated systematic errors

Additional Cases Covered in V&V 20

V&V 20 Case 4



Case 4 considers a combustion flow with the validation variable being duct wall heat flux q at a given location. The experimental q is inferred from temperature-time measurements at the outside combustor duct wall using a data reduction equation that is itself a model. The predicted q is from a simulation using a turbulent chemically-reacting flow code to model the flow through the duct.

Interpretation of Validation Results with No Assumptions Made about the Error Distributions

$$\delta_{\text{model}} = E - (\delta_{\text{input}} + \delta_{\text{num}} - \delta_D)$$

If $|E| \gg u_{\text{val}}$
then probably $\delta_{\text{model}} \approx E$.

If $|E| \leq u_{\text{val}}$
then probably δ_{model} is of the same order as or less than $(\delta_{\text{num}} + \delta_{\text{input}} - \delta_D)$.

Interpretation of Validation Results with Assumptions Made about the Error Distributions

$$\delta_{\text{model}} = E - (\delta_{\text{input}} + \delta_{\text{num}} - \delta_{\text{D}})$$

In order to estimate an interval within which δ_{model} falls with a given probability or degree of confidence, an assumption about the probability distribution of the error combination $(\delta_{\text{input}} + \delta_{\text{num}} - \delta_{\text{D}})$ must be made. This then allows the choice of a coverage factor k such that

$$U_{\%} = k_{\%} U_{\text{val}}$$

One can say, for instance, that $(E \pm k_{95} U_{\text{val}})$ then defines an interval within which δ_{model} falls about 95 times out of 100 (i.e., with 95% confidence) when the coverage factor has been chosen for a level of confidence of 95%.

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Questions?