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Research Article

GRS Method for Uncertainty and Sensitivity Evaluation of Code Results and Applications

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Abstract

During the recent years, an increasing interest in computational re-evaluation model calculations by best estimate calculations supplemented by uncertainty analysis has been observed. The evaluation of the margin to acceptance criteria, for example, based on the upper limit of the calculated uncertainty range. Uncertainty analysis is required to be obtained from “best estimate” thermal-hydraulic code results. The accuracy would be presented for comparison with regulatory acceptance criteria. The results are presented to quantify the uncertainty of computer code results. The results are presented together with applications to a large break loss of coolant accident simulation simulating containment behaviour.

1. Introduction

Best estimate computer codes are used to calculate postulated accident scenarios in a best estimate way and not in a conservative way. There is an increasing interest in the use of best estimate model calculations by best estimate calculations supplemented by uncertainty analysis. The USA Code of Federal Regulation (CFR) 101.61 requires the use of best estimate code plus identification and quantification of uncertainty.

conservative computer code models listed in Appendix K of the CFF

Code predictions are uncertain due to several sources of uncertainty in plant and fuel parameters. These uncertainties, for example, come from modelling, variation and imprecise knowledge of initial and boundary conditions developed based on experiments which can simulate the complex conditions in a simplified way only. Most of the experiments are conducted under conditions that are not representative of actual conditions. Uncertainty due to imprecise knowledge of parameter values in computer code models and their distributions. These distributions should be taken into account for i

Stochastic variability due to possible component failures of the system during the analysis. The single failure criterion is still taken into account in the safety analysis and requirements of redundancy. The probability distributions of component failures, not of demonstrating the effectiveness of emergency cor

The aim of the uncertainty analysis is at first to identify and quantify the sources of uncertainty. Their propagation through computer code calculations provides the uncertainty in the results. The evaluation of the margin to acceptance criteria, for example, should be based on the upper limit of this distribution for the code results. A safety analysis is needed if useful conclusions with regard to prediction accuracy are to be obtained from “best estimate” thermal-hydraulic code results. The accuracy would be presented for comparison with limits for accepta

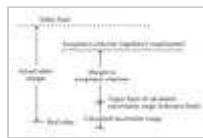


Figure 1: Margin illustration.

Section 2 describes the GRS method, Section 3 presents example results, and Section 4 provides conclusions.

2. Description of the GRS Method

Among others, GRS method [2] has been developed for the detection and quantification of uncertainty. The method is described by ranges and probability distributions. For the analysis of computer code results, a number of calculation runs, all identified uncertain parameters are varied. These include uncertain input values, models, initial and boundary conditions, maximum time step size, and so forth. Model uncertainties are expressed as additive or multiplicative terms, or by a set of alternative model formulations. Uncertainties in the input parameters, model phenomena, are to be taken into account in the code validation procedure. Code validation results are compared with acceptance criteria and uncertainties.

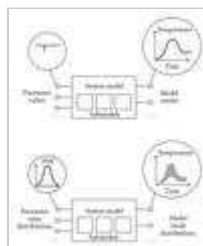


Figure 2: Consideration of input parameter variation method.

The selection of parameter values according to their specified probability distribution for the evaluation of the calculation results requires a method. Following a set of statistical techniques. The advantage of using these techniques is that the number of calculations needed is independent of the number of uncertain parameters. In order to quantify the effect of these parameters, they are varied simultaneously. In order to quantify the effect of these parameters, they are varied simultaneously. Because the number of calculations is independent of the number of uncertain parameters, the number of input parameters is necessary to reduce their number in order to make the analysis as described later.

The number of code calculations depends on the requested probability tolerance limits used in the uncertainty statements of the results. The number of calculation runs is given by Wilks' formula [3, 4], for example $b \times 100$ is the confidence level (%) that the maximum code result is less than or equal to the b th percentile of the corresponding output distribution, which is obtained from a random sample of limited size. The formula is: $1 - a^n - n(1 - a)a^{n-1} \geq b$. The minimum number of calculations is given by n .



Table 1: Minimum number of calculations n for various confidence levels b and tolerance limits a .

The probabilistic treatment of parameter uncertainties allows quantification of the uncertainty range, the knowledge is expressed in terms of probability distributions. This interpretation of probability is used for a parameter with a known value. The classical interpretation of probability as the limit due to stochastic variability, is not applicable here.

The probability distribution can express that some values in the range are more appropriate parameter value than others. In the case that no preference is specified, that is, each value between minimum and maximum is equally likely. As the consequence of this specification of probability distribution, the results also show a probability distribution, from which uncertainty limits can be determined.

A total number of n code runs are performed varying simultaneously according to their distribution. The n values of the considered output are ordered from largest to smallest, $Y(n)$. Therefore, the name-order statistics is used for Wilks' formula. A 5th percentile value with a confidence level of 95% is obtained by selecting $Y(1)$ limit, for example. A 5th percentile value with a confidence level of 95% (95% / 95%) two-sided tolerance limit is obtained by selecting $Y(1)$.

Another important feature of the method is that one can evaluate the parameter uncertainties for the uncertainties of the results. These results provide information as to where to improve the statistical analysis, where to improve the modelling, where to improve the standardised rank regression coefficients, rank correlation coefficients, and so forth, with uncertainty. The difference to other known uncertainty methods, the analysis and not of prior estimates and judgements. This procedure (PIRT) by extensive expert staff-hours in [5] is known to be effective. Measures are available simultaneously for all single-valued (e.g. point estimates) and valued (time dependent) output quantities of interest. The method

using approximations like fitted response surfaces. Similar method uncertainty method is presented in [6].

The different steps of the uncertainty analysis according to the GRS uncertainty and sensitivity analyses (SUSA) developed by GRS [7] are applied during the uncertainty and sensitivity analysis.

3. Applications

The GRS method for uncertainty and sensitivity evaluation of applications have been performed in GRS to investigate loss of systems of pressurised water reactors, as well as related experiment hydraulic computer code ATHLET. Another uncertainty and sensitivity experiment simulating containment behaviour using the computer

3.1. Thermal-Hydraulic Applications Using the ATHLET Computer

Several uncertainty and sensitivity analyses were performed by ATHLET simulating breaks of the primary and secondary side coolir

- (i) separate effects experiment OMEGA heater rod bundle Te
- (ii) integral experiment LSTF-CL-18, 5% cold leg break, accu
- (iii) PWR 5% cold leg break, accumulator injection into hot le
- (iv) integral experiment LOFT L2-5, $2 \times 100\%$ cold leg break,
- (v) PWR $2 \times 100\%$ cold leg break, combined ECC injection int
- (vi) PWR 10% steam line break,
- (vii) PSB-VVER 11% upper plenum break experiment, UP-11-(

One out of these applications is described in the following section.

3.2. Application to a German PWR Reference Reactor, 2×100

A double ended cold leg offset shear break design basis accident investigated. The fuel rod peak linear heat generation rate is 53 assumed. ECC injection is into cold and hot legs. The accumulator the primary system below a pressure of 2.6 MPa. High- and low-p assumed in the broken loop check valve for ECC injection from acc hot leg accumulator is unavailable due to preventive maintenance. unavailability, agreed between applicants and assessors.

The uncertainty analysis considered 56 uncertain input paramet parameters to select different model correlations for heat transfer reactor vessel, 1 for temperature of accumulator water, 1 for c distribution in the core, 1 for hot channel factor, 5 for gap width (and 2 for convergence criteria. The model parameters comprise cri wall and interfacial shear, form loss, main coolant pump head, and

A total number of 100 calculations were performed using the code

3.3. Maximum Clad Temperature

Figure 3 shows at any point of time, at least 95% of the combin

calculated clad temperatures is below the presented uncertainty limit of at least 95%. For each instant of time, the desired tolerance results. A “conservative” calculation result is shown for comparison with default values of the models and conservative values for the initial heat, gap width of fuel rods between fuel and clad, fuel pellet thermal conductivity, and the water. All these conservative values were also included in the distribution analysis. The maximum clad temperature of the conservative calculation is within the tolerance limits of the uncertainty analysis over the whole transient. The regulatory acceptance criterion for peak clad temperature is 1200 °C.

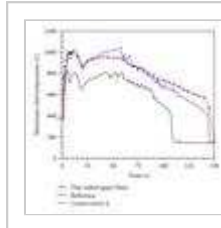


Figure 3: Calculated one-sided 95% / 95% calculation compared with a “conservative” calculation for a reactor during a postulated double ended offsite event.

The “conservative” calculation is representative for the use of initial and boundary conditions. Such an evaluation is possible in the USA. The uncertainty of code models is not taken into account. Boundary conditions will bound these model uncertainties. That is the present example. An uncertainty analysis quantifies uncertain input parameters and uncertainties. The peak clad temperatures, however, are bounded by sensitive parameters gap width and pellet thermal conductivity. The extent of conservatism implemented in the conservative calculation [1] requires that “uncertainties in the analysis method and input parameters uncertainty in the calculated results can be estimated” when a best estimate calculation is used.

According to the US Code of Federal Regulations, Title 10, Section 50.103, conservative models to be applied in conformity with the requirements of “ECCS Evaluation Models” of the Federal Regulations [1]. This margin to licensing criteria is available by changing from conservative to uncertainty analysis.

The confidence level 95% denominates that the 95th percentile is provided by a (95%, 95%) statement. This conservatism is the real statement by a conservative calculation is not needed. Methods for comparison and quantification of “conservatism” and additional statistical test proving that the conservative calculation is

3.3.1. Sensitivity Measures

Sensitivity measures indicate the influence of the uncertainty in input parameters on the Spearman rank correlation coefficient is used as sensitivity measure. The sensitivity of the respective input parameter uncertainty on the final clad temperature during the blowdown phase; see Figure 4. The sensitivity measure gives the direction and magnitude of deviations when the input uncertainty varies by one standard deviation. Positive sign means that input parameter value and result tend to increase together. Negative sign means that input parameter value tends to increase the clad temperature and the result tends to move in opposite direction to decrease the clad temperature and vice versa.

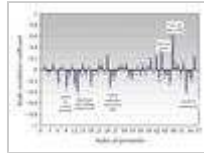


Figure 4: Sensitivity measures of the blowdown input parameters (rank correlation coefficient)

The most important parameter uncertainties, out of 56 identified the blowdown peak clad temperature uncertainty are

- (i) fuel rod gap width for low burn up (positive sign),
- (ii) fuel heat conductivity (negative sign),
- (iii) minimum film boiling temperature (negative sign),
- (iv) model for critical heat flux (negative sign: Biasi correlated change from nucleate to transition boiling compared to the Her
- (v) reactor initial power (positive sign),
- (vi) 2-phase multiplier in horizontal pipe (negative sign: high \Rightarrow higher water content in core due to lower break flow \Rightarrow lower

The most important parameters for the peak clad temperature unc

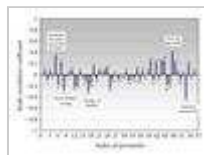


Figure 5: Sensitivity measures of the reflood input parameters (rank correlation coefficient)

- (i) fuel heat conductivity (negative sign),
- (ii) fuel rod gap width for low burn up (positive sign),
- (iii) model for 1-phase convection to steam (positive sign, i. temperatures than Dittus-Boelter II),
- (iv) number of droplets (negative sign: number of droplets h
- (v) steam-droplet cooling (negative sign: higher cooling ten

3.4. Application to the Experiment HDR T31.5 Simulating Co

The experiment T31.5 on the HDR containment facility simulates steam and gas release into the containment according to the low pressure phase was performed with emphasis on pressure buildup in the hydrogen distribution was measured during a long term phase of mixture were injected.

A total number of 200 calculations were performed using the code influence of all considered uncertainties on the calculated pressure of time is shown in Figure 6. A total of 79 uncertain parameters were experimental facility, initial and boundary conditions.

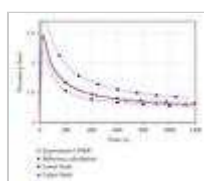


Figure 6: 95% / 95% uncertainty interval, pressure in the upper part of the containment

Sensitivity measures about the influence of the uncertainty in input parameters on the HDR containment versus time are presented in Figure 7. W versus time on the maximum pressure. Decreasing influence with decreasing convection for

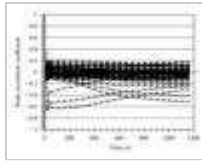


Figure 7: Sensitivity measures for pressure in

- (i) free convection, parameter 72, negative sign,
- (ii) forced convection, parameter 73, negative sign,
- (iii) condensation at wall, parameter 74, negative sign.

Increasing with time are the following parameters because of decrease

- (i) thickness of liner, parameter 79, negative sign,
- (ii) surface of liner, parameter 77, negative sign,
- (iii) heat capacity of concrete structures, parameter 69, negative sign.

4. Conclusions

Two applications of the uncertainty method proposed by GRS methodology is that no a priori reduction in the number of uncertainty screening calculations is necessary to limit the calculation effort. A reduction in the number of calculations is achieved in the uncertainty analysis. The method accounts for the combined effect of the quantified uncertainties. This would be difficult or even impossible to achieve by a traditional method or transients.

The number of calculations needed is independent of the number of parameters in the uncertainty analysis. It does, however, depend on the requested tolerance (percentile) of the combined effect of the quantified uncertainties, results. The tolerance limits can be used for quantitative statements.

Another important feature of the method is that it provides sensitivity measures for all input parameter uncertainties on the results. The measures permit an analysis of the impact of each parameter and provides guidance as to where to improve the state of knowledge, where to improve the modelling of the computer code, or where to improve the ranking is a result of the analysis and its inputs and not of an arbitrary choice. Sensitivity measures are available simultaneously for all single and multiple output quantities as continuous valued (time dependent) output quantities of interest without the use of approximations like fitted response surfaces. The method is used in different applications by various international institutions in the nuclear industry.

A challenge in performing uncertainty analyses is the specification of the uncertainty distributions for the input parameters. Investigations are underway to transform data measured from a thermal-hydraulic model parameters with uncertainties. Care must be taken to use analytical information to specify uncertainty distributions. This is a challenge for all uncertainty methods.

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