

Overview of Draft Technical Letter Report on “Important Aspects of Probabilistic Fracture Mechanics (PFM) Analyses”

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Development of Guidance for Probabilistic Fracture Mechanics (PFM)

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Background



- First deliverable of the PFM Regulatory Guide development project
- GOALS
 - Create a technical report within a relatively short period of time to highlight what NRC views as important concepts for PFM regulatory applications
 - Means to communicate NRC's initial thoughts on PFM, prior to the development of a Regulatory Guide and supporting NUREG
 - Means to generate discussion and obtain feedback

Technical Letter Report Outline



1. Introduction: motivation and objectives
2. Definition of PFM, including similarities and differences with deterministic analyses
3. Analysis models
4. Analysis inputs
5. Uncertainty framework
6. Analysis outputs

Important Definitions

- Accuracy and Precision
- Aleatory Uncertainty
- Epistemic Uncertainty
- Probabilistic Fracture Mechanics
- Probabilistic Risk Assessment
- Realization
- Sensitivity Analysis
- Sensitivity Study
- Stability Analysis
- Uncertainty Analysis
- Uncertainty Propagation
- Validation and Verification

1- Introduction: Motivation for Increased Confidence in PFM and Primary Objective

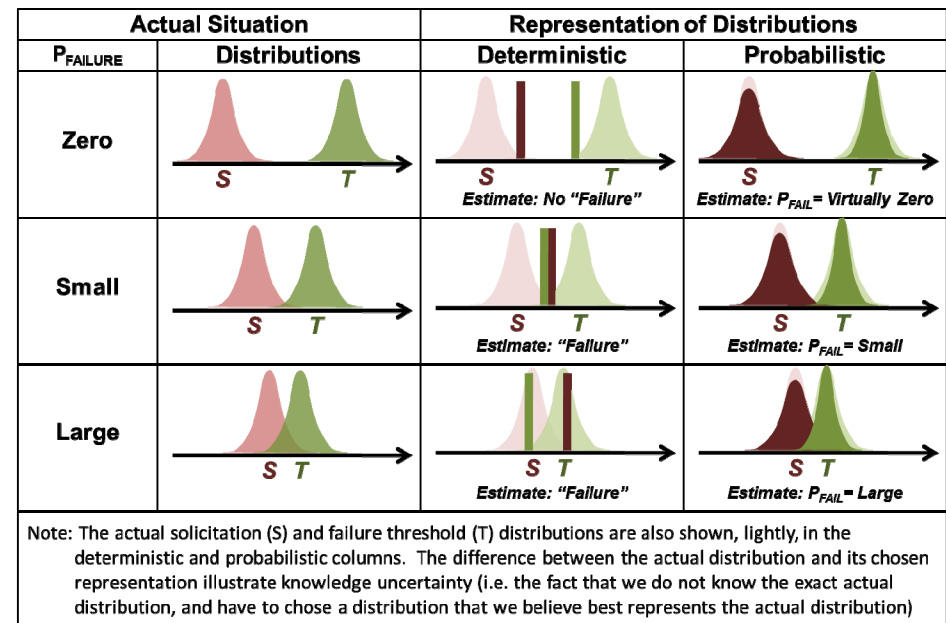
- Regulating the use of structural nuclear materials:
 - Before 1995: deterministic approaches (safety factors, margins, conservatisms...)
 - Since PRA policy statement in 1995: towards risk-informed and performance-based regulations
 - Fracture behavior at heart of safety of nuclear structures
 - Consensus Codes and Standards typically deterministic conservative approaches to bound uncertainties
 - PFM can complement deterministic analyses and quantify uncertainties
 - PFM pros and cons:
 - Direct representation of uncertainties through the use of best-estimate models and distributed inputs
 - Permits determination of the direct impact of uncertainties on the results, identification of problem drivers
 - Often more complex, requires more data to construct distributions, more time consuming
 - NRC experiences with PFM regulatory reviews overall challenging
 - Some successes (risk-informed inservice inspection, elimination of vessel weld inspections)
 - Many difficulties: input choice justification, 'black-box' tools, insufficient V&V, sensitivity analyses & studies
- **Primary objective: develop a methodology that is sufficiently general to be suitable to guide the development and/or critique of any PFM analysis, and that also effectively and logically manages the level of detail and breadth of scope that PFM analyses can take on**

2- Definition of a PFM Analysis: Generalities on Fracture Mechanics

- Basic principle: is $K_{\text{Applied}} [\leq \text{ or } >] K_{\text{Resistance}}$?
 - $K \sim \sigma\sqrt{a}$
- Six variables comprise the most important aspects of any DFM or PFM analysis
 - K , σ , a , and component geometry are common to any fracture mechanics analysis
 - Environment and time also required to completely understand, characterize, and evaluate virtually any fracture mechanics analysis
- Human interaction (inspection) is considered as a measure to confirm the validity of the fracture mechanics model
- Repair or mitigation is considered as a measure to change some or all key variables
- Accurate representation of interdependences in the DFM or PFM model is key to the development of a model that reasonably approximates reality
- Multiple time and spatial scales may need to be modeled

2- Definition of a PFM Analysis: Deterministic vs. Probabilistic

- K_{Applied} and $K_{\text{Resistance}}$ inherently distributed
- DFM and PFM are fundamentally similar
 - Mathematical abstractions
 - Common goal of representing uncertainties in mathematical form for problem solution
 - Not possible to account for every uncertainty
- PFM analyses will represent most, but not necessarily all, uncertainties as important distributed quantities
- Both PFM and DFM use specific metrics to quantify results
- Use of DFM or PFM methods does not represent an either-or choice
- DFM and PFM methods can, and have, been used as complementary parts of a safety case
- Failure *probability* is one possible numeric outcome of a PFM analysis: impediment to the acceptance of PFM techniques



2- Definition of a PFM Analysis: Considerations on Use of PFM

- Reasons for shift from deterministic to probabilistic approaches in regulating component integrity
 - PRA policy statement in 1995
 - Factors unanticipated in the design phase, and/or not addressed by Codes and Standards
 - New degradation mechanisms or deficiencies discovered after design and construction: cost benefit analysis
 - PFM ideally suited to calculate perform risk assessment, required by backfit rule
 - Unanticipated service occurrences not covered by Codes and Standards cannot easily be accounted for in standard deterministic demonstrations
 - PFM potentially more adaptable
 - DFM and PFM require continuous updating
 - Use of structures beyond their design, or licensed, lifetimes
 - Application of large safety factors may become impracticable
 - PFM can better quantify the risks associated with operation of NPPs beyond the initial licensed life
- Other benefits
 - Rigor of process needed to build models can revealed deficiencies previously overlooked
 - Force analysts to be more critical of data used to develop models or inputs

3- Models for PFM Analyses: Model Selection

- Collection of deterministic models linked in a probabilistic framework
- Considerations for model selection
 - Best-available representation
 - Engineering judgement
 - Accuracy and precision vs. mathematical/computational feasibility
 - Documentation of choices, proper justification, discussion of model uncertainties/biases
- Different ‘types’ of models
 - Mathematically derived models with various degrees of theoretical underpinning
 - Empirical models that rely only on data fitting (lab vs. field data)
 - Importance of data quality and relevancy
 - Statistical characterization of model uncertainties and biases
 - Computational models
- Model validity considerations
 - Applicability bounds: underlying data, expert judgement, physical limits of system
 - Importance of documentation
 - How does the code/model deal with sampled inputs that are out of bounds?

3- Models for PFM Analyses: Gaining Confidence in Models

- Heavy reliance on documentation of models and on V&V activities
- Model V&V
 - Verification: QA process by which the coded model is verified to meet the defined software requirements, i.e. the mathematics of the model are correctly coded
 - Independent verification
 - Validation: QA process to assess degree to which the chosen model represents the physical system
 - Does the model predict expected values, trends, and actual data?
- Documentation of model uncertainties
 - Simplifying assumptions, scatter in dataset used to create model, lack of data, simplified models for computational reasons
 - Identification of sources of uncertainty, qualification/quantification of uncertainty
 - Assurance that limitations of PFM code are understood and accounted for
- Consideration of alternative models
 - Potential sensitivity studies
 - Strengthen argument for choice of a particular model

4- Inputs for PFM Analyses: Uncertainty Type, Distributions, Bounds

- Input uncertainty classification: constant vs. random
- Random inputs are usually represented by mathematical distributions
 - Sampling based methods use probability distributions
 - Multitude of distribution types
 - Continuous or discrete distributions
 - Other non-sampling based methods exist
- Construction of input distributions
 - Availability and pedigree of data
 - Distribution type and fitting distribution parameters
 - Distribution bounds, skewness, kurtosis
 - Goodness-of-fit tests, engineering judgement
- Input bounds
 - Many engineering applications require bounds to remain within physical reality
 - May choose bounds to limit inputs to a model's domain of applicability
 - Important to exercise caution and justify choice input bounds
 - Different models may have different validity domains
 - True physical bounds not always well known

4- Inputs for PFM Analyses: Assumptions, Conservatisms, Dependencies

- Justification of input assumptions (technical basis for inputs)
 - Need stronger technical bases for most influential inputs
 - May not need very strong basis for less influential inputs
 - Scrutinize important input distribution tails for low probability events (sensitivity studies)
 - If large impact of distribution tails, may need more data, expert elicitation, more refined statistics...
- Conservatisms in inputs and models may not be appropriate in PFM
 - Looking for best-estimate with quantified uncertainty
 - Should attempt to quantify known conservatisms, and understand their impact
 - Conservatism could be applied on final criteria instead of inputs or models
- Input variables usually assumed to be independent
 - Known variable dependencies should be modeled
 - Avoid underestimation of input variable importance (sensitivity analysis)

5- Uncertainty Characterization: Aleatory and Epistemic

- Definitions
 - Aleatory Uncertainty: (perceived) natural, unpredictable variation in the performance of the system under study over which there is little control, or inherent randomness in the future
 - Epistemic Uncertainty: due to a lack of knowledge about the behavior of the system that is conceptually resolvable
- Most inputs contain both types of uncertainty, characterization is not absolute
- Aleatory and epistemic uncertainties can sometimes be separated
 - Typical means of separation: double nested loop
 - Can provide additional insights: confidence intervals on quantities of interest
 - Potentially higher computational cost
 - Not all codes have the capability to separate aleatory and epistemic uncertainty
- Sensitivity studies consisting of changing the uncertainty characterization can be helpful to remove some of the most influential variables from the sensitivity analysis
 - Gain insights on second-order importance of inputs

5- Uncertainty Characterization: Representation and Propagation

- Choice of random variables often responsibility of experts
 - Need capture major sources of uncertainty
 - Need to have proper representation of uncertainty (distributions)
 - Potentially need to separate epistemic from aleatory uncertainty
- Context of analysis (specific plant system versus generalization) may have important impact on choice of distributions and output uncertainty
- Several possible probabilistic frameworks to represent and propagate uncertainty
 - First/Second Order Reliability Method good for well defined response surfaces
 - Sampling based techniques are more general but more ‘expensive’
 - Sample random variables at each realization
 - Simulate the system for each sample
 - Collect results of independent realizations and combine into probability distributions of possible outcomes
- Many Monte Carlo sampling techniques
 - Simple Random
 - Latin Hypercube
 - Importance Sampling
 - Adaptive Sampling

5- PFM Framework/Code: Verification and Validation

- High confidence in PFM software and analyses requires verification and validation
 - Verification establishes the correspondence between the PFM computer code and its specifications
 - Validation establishes whether the PFM code is fit for its intended purpose
- Quality Assurance (QA) requirement of 10CFR50 Appendix B
 - Documented requirements for the PFM software, including development, procurement, maintenance, testing and configuration management
 - Criteria for QA plans: ASME NQA-1 or NUREG/BR-0167
- V&V of both deterministic sub-models and probabilistic framework is required
 - Validation can be done against data, analytical solutions, or output from another validated tool (e.g. FEA)
 - Validation required for each output, and for overall solution
 - Validation may be application-specific
 - Graded approach useful (most important quantities require higher degree of validation)
 - For low probability events with scarce data, sensitivity studies and study of trends is useful
- QA and V&V extent can vary based on...
 - Application permanence in time, safety significance
 - Scale and complexity of code and models
 - Pre-verified software vs. one-time-use
 - Novel code vs. accepted consensus
- Other factors that increase confidence: peer-review, detailed documentation, availability of source code, benchmarking

6- PFM Outputs: Convergence and Uncertainty Analysis

- Reliable, realistic PFM results require more than a single PFM analysis with a given PFM code
 - Should conduct multiple runs to demonstrate that solution is converged, uncertainties are properly represented, and parameters driving the problem have been sufficiently characterized
- Convergence is often a matter of perspective, linked to confidence in decisionmaking
 - Need to define a target threshold value or range for pass/fail criterion
 - A solution is considered converged enough if the uncertainty is low enough that it does not change the conclusion that could be drawn from the analysis
 - Acceptable level of uncertainty decreases when getting closer to the threshold value
 - Mean may converge fast, 99th percentile no so fast...
- Temporal, spatial, and statistical converge need to be demonstrated
 - Temporal and spatial: time stepping and discretization
 - Statistical: stability analysis: changes in sample size, random seed, sampling technique, etc.
- With convergence achieved, uncertainty analysis on outputs of interest can be performed
 - Construct distributions of outputs/quantities of interest (QoI)
 - Many possible representations (probability vs. time, distribution of spatial location, horsetail plot of all realizations, PDF or CDF of a QoI, etc.)
 - QoI may be mean or percentile

6- PFM Outputs: Sensitivity Analysis

- Sensitivity analyses: understanding the relationship between problem input and output uncertainties
 - Useful to identify drivers of the problem being analyzed
 - Important in developing sampling options which provide converged results
- Two main categories of sensitivity analyses:
 - Graphical qualitative methods: scatterplots
 - Can deal with any set of arbitrarily-distributed input and output variables
 - Direct qualitative visual indication of sensitivity
 - Identification of good candidates for importance sampling
 - Quantitative ‘variance-explained’ methods: regression analysis
 - Several methods should be combined to eliminate false positives
 - Goal: show that the final results are comprehensive, well supported, and well documented
 - Common methods: linear, rank, quadratic, or Gaussian regression; regression trees, and multivariate adaptive regression splines (MARS)

6- PFM Outputs: Sensitivity Studies

- Sensitivity studies: exploring the specific influence of important parameters and analyzing specific scenarios of interest
 - Useful to better understand the physics of a problem
 - Used to perform “what if” analyses, to revisit some assumptions, and to investigate alternative scenarios in support of the defense in depth
- Two main categories of sensitivity analyses:
 - Deterministic sensitivity studies
 - Change one or more constant parameters that could influence the output one-at-a-time (alternative scenarios)
 - Usually focus on the physics of the system
 - Build confidence on the theoretical aspect of the analysis
 - Probabilistic sensitivity studies
 - Change one or more random input distributions: shift median, change uncertainty (kurtosis), change focus (skewness)
 - Change uncertainty classification between aleatory and epistemic
 - Focuses on the changes in probabilities of occurrence associated with each change

6- PFM Outputs: Problem Drivers and Confidence Demonstration

- Sensitivity analyses and studies should be performed to identify the drivers of the problem, and revisit the uncertainty of those drivers, to gain confidence in PFM conclusions
 - Important to ensure that the distributions for problem drivers are representative
 - Just as important to study the consequences if these distributions are not representative
- The parts of a given input distribution that drive the answer to a problem should be identified and sufficiently sampled
 - Role of importance sampling if distribution tails are important
- When confidence cannot be gained in the chosen distribution for a given parameter that drives the problem, several alternatives can be explored
 - Expert elicitation
 - Additional data collection

Future Work

- This Technical Letter Report is only the first step in the PFM Regulatory Guidance development project
- Next phase:
 - Development of a Draft Guide on PFM
 - Development of a detailed technical basis NUREG document
- In parallel: pilot study to test the Draft Guidance
 - Identify deficiencies and areas that need improvement
 - Lessons-learned will be incorporated in final Draft Guide document, and the technical basis NUREG will be revised accordingly