

# Future, Model, and Parameter Uncertainty

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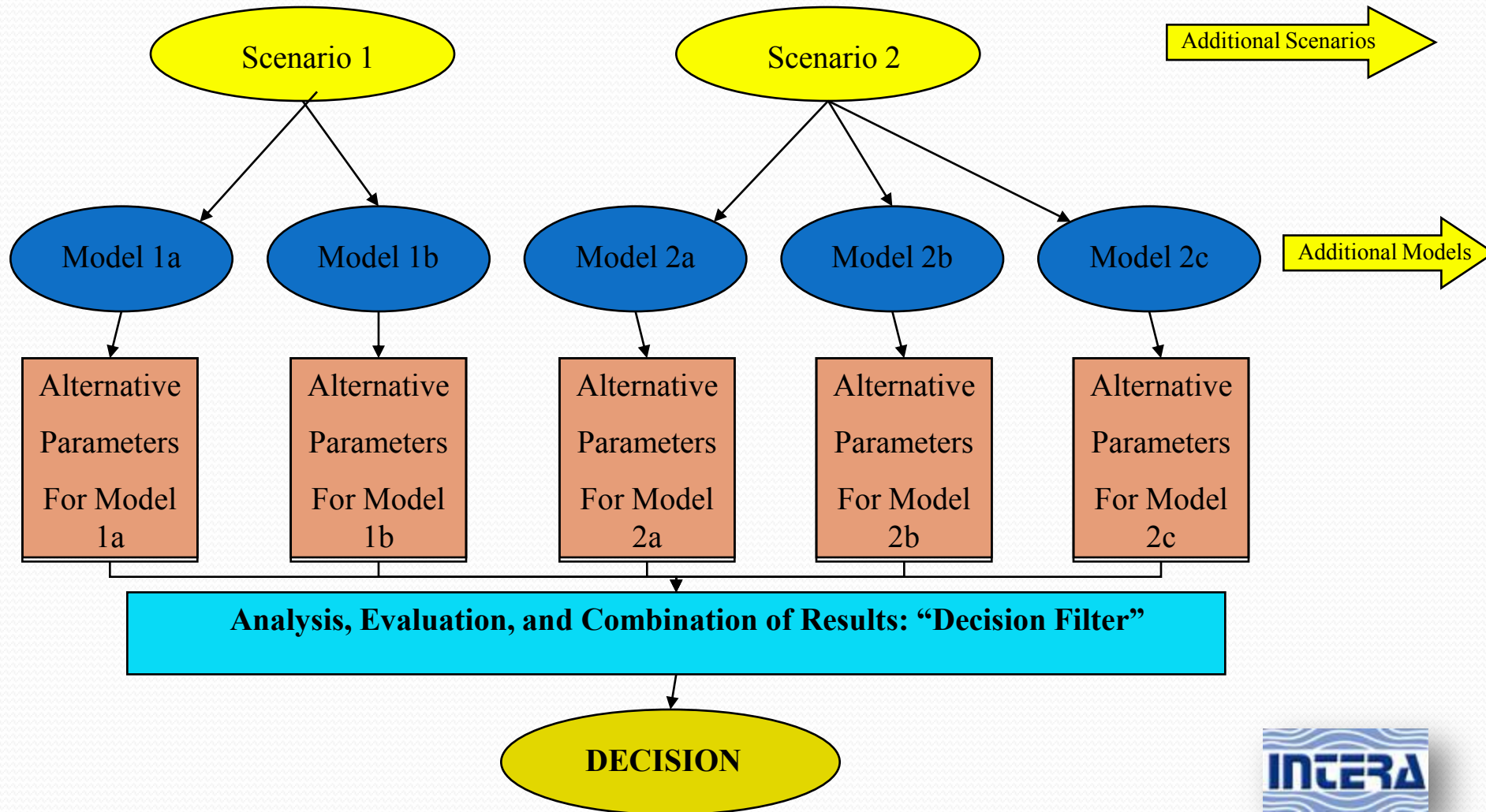
# Scope of the Presentation

- Structure of uncertainty analysis
- Characterization of uncertainties
- Propagation of uncertainties
- Implications for results of performance assessments

# Introduction

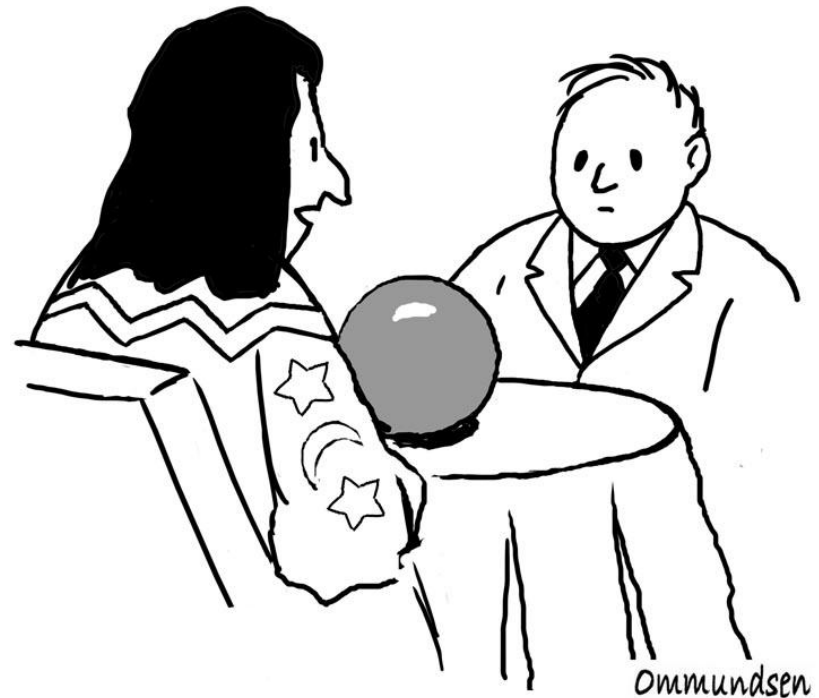
- Sensitivity and uncertainty analyses are well-studied in the literature
- Caution must be used in adopting available methods for use in performance assessment
  - Performance assessment is an unusual activity
  - Frequent misunderstandings have arisen
- NCRP Report 152 introduced a new term to describe the approach for safety assessment
  - Importance Analysis: the integration and interpretation of performance assessment results to identify those that influence the decision regarding compliance

# Structure of Uncertainty or Importance Analysis



# A General Approach for Treating Uncertainty

- Evaluate Multiple Line of Reasoning for Each Type of Uncertainty
  - Consideration of alternative scenarios for future uncertainty
  - Consideration of alternative models for conceptual uncertainty
  - Consideration of alternative parameter sets for parameter uncertainty
- The result is a potentially large number of calculations that represents the uncertainty



**“Is this needed for a Bayesian analysis?”**

# Decision Filter

- Each set of scenario, model, and parameter set is assigned a weighting factor
  - May be implicit (disregarding a model = weighting factor of zero)
  - May be qualitative (Model 1a is better than Model 1b)
  - May be quantitative (probabilistic)
- The filter defines how the information is used in making a decision
- The choice of filter depends on
  - Assessment context
  - Philosophy of analyst
- A few comments about probabilistic approaches
  - Subjective probabilities and ranges are easy to assign: we are not representing variability
  - Technically superior way to span the range of the input space
  - That superiority comes at a cost

# Characterization of uncertainties 1

## Aleatory vs. epistemic

- Performance assessment uncertainties are dominated by epistemic uncertainties (Type B)
- Even when large amounts of data exist, uncertainty about application to future field conditions is more important
  - Transition to different constitutive behavior under different boundary conditions (e.g. hydraulic conductivity)
  - Transition to different constitutive behavior in time
- Aleatory uncertainties (Type A) are generally unimportant
  - This situation differs from power plant risk assessments
  - Also differs from other types of risk assessment activities

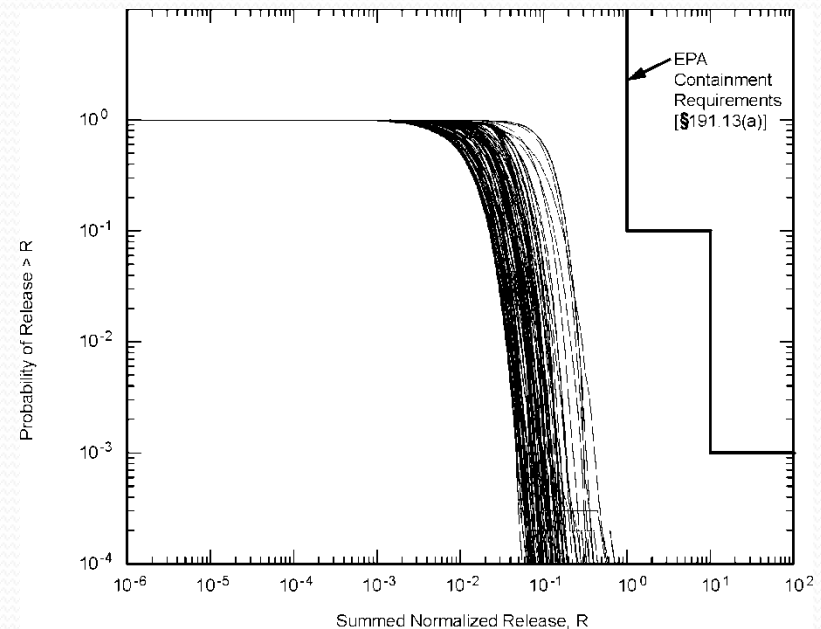


Figure 6-35. Distribution of CCDFs for Normalized Radionuclide Releases to the Accessible Environment from the WIPP, Replicate 1

**Attempted differentiation between Type A and Type B uncertainties in WIPP performance assessments**

# Characterization of uncertainties 2

## Future, Model, and Parameter

- When is an uncertainty a “scenario” a “model” or a “parameter”
  - Largely semantics; these divisions are not fundamental
  - One approach: does the initiating FEP act on the system or is it part of the system?
- More important: is the issue clearly addressed in the assessment?



# Characterization of Uncertainties 3:

## Features, Events, and Processes (FEPs)

- Features
  - Aspects of the disposal system associated with performance
  - Generally thought of as physical components
- Events
  - Discrete occurrences
  - Relatively short duration
- Processes
  - Longer term evolutionary aspects of the system
  - Generally represent relationships between features

In practice, little differentiation between these three, and one simply discusses “FEPs”



# FEPs Background

- Scenario approaches developed in the 1980s
  - Sandia methodology
  - Developed for U.S. HLW waste program
  - Legal requirement to represent all events and processes
  - Requirement to combine them probabilistically
  - Intended to identify scenarios
- Scenario approaches developed in the 1990s
  - SKB methodology
  - A move away from probabilistic approaches
  - Inclusion of FEPs representing the model
- Scenario approaches developed in the 2000s
  - Multiple methodologies with common features
  - Extension to FEPs for near surface disposal

# Scenario Development Methodologies

## four basic steps

- Comprehensive FEP list
- Screening
- Describing relationships between FEPs
- Arranging them into calculational cases, or scenarios

Differences between published approaches represent differences in ordering of these basic steps

## Why do we use FEPs?

- The historical (1980s) use was to identify all initiating events and processes for scenarios
- Modern usage is broader, and includes both identification of scenarios and construction of models
- The path from FEPs to models is not clear
  - Typically a leap occurs between FEPs and models
  - Current assessments often receive criticism for this leap
  - The reality is that models are developed using professional judgment, informed by FEPs
- FEPs are best viewed as a communication tool, not a fundamental feature of scenarios and models
- Strong use as an auditing tool to ensure conceptual completeness

# Conceptual model uncertainty:

## Origins of alternative conceptual models

- Differing assessment context
  - Degrees of conservatism
  - Regulator vs. developer
  - Differing analysts
- Differing scenario definitions
- Exploratory conceptual models
  - Evaluation of alternative assumptions
  - Performance margin analysis

# Conceptual mode uncertainty

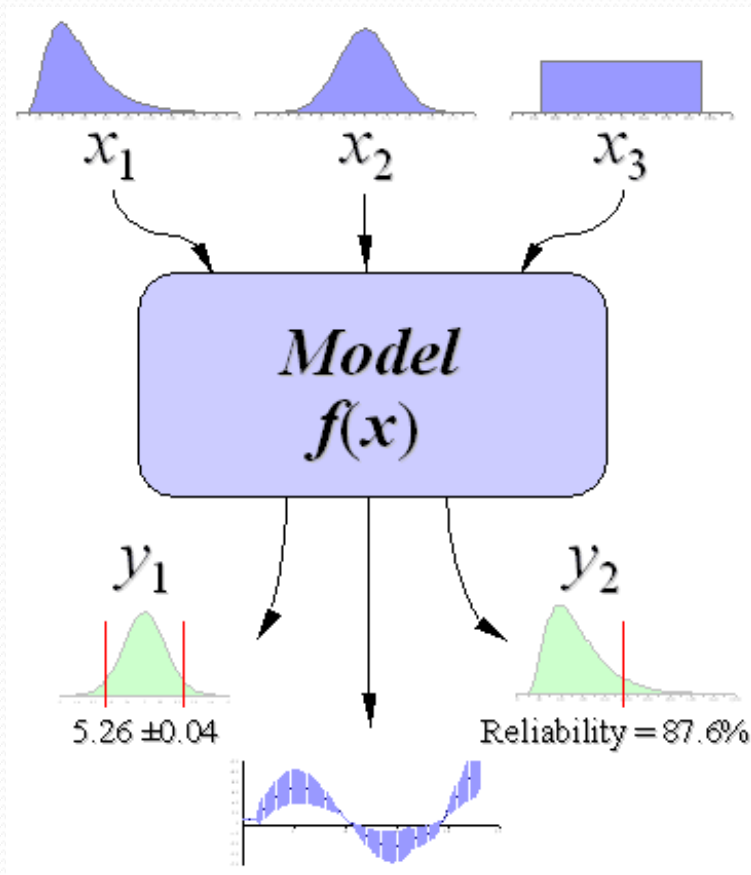
## Performance margin analysis

- Evaluate assumptions that are credible, but difficult to defend
- A quantification of “conservatism”
- Can provide strength to licensing arguments

# Resolution of alternative models

- Data are often not available
- Elimination of a competing model should be based on evidence
- Model intercomparison seen as a primary tool for producing credibility
- Necessarily involves consideration of alternative points of view
- Alternative models are best resolved by focusing on details of each to come to consensus

# Probabilistic Treatment of Parameters



- Sample from input distributions
  - Random sampling
  - Latin Hypercube Sampling (LHS)
- Run the model for “enough” times to produce a stable output distribution
  - There are no useful rules for establishing stability
  - Large numbers of realizations usually needed
- Need to manage the massive inputs and outputs



# Non-probabilistic Treatment of Parameter Uncertainty

- Easier to communicate and understand
- May be harder to defend
- Needs to address the same uncertainties

# Maximum entropy approach to parameter distributions

If you know...	The distribution should be...
Range (a,b)	Uniform distribution on (a,b)
Mean and standard deviation ( $\mu, \sigma$ )	Normal distribution
Positive and given mean ( $\mu$ )	Exponential
Mean, standard deviation, range ( $\mu, \sigma, a, b$ )	Beta
Mean occurrence rate between discrete events	Poisson

When data span many orders of magnitude, often a log-uniform or log-normal distribution is used.

# Summary

- Performance assessment is a practical analysis
  - Do not become bound up with semantics
  - Need to identify uncertainties that truly represent a concern
  - Different people see “concern” in different ways
- Any information is useful **if viewed in the right context**
- Clarity in communication is needed
  - A mixture of probabilistic and nonprobabilistic
  - Clear reasoning on the reason specific scenarios and models are used, and what they are intended to represent