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# The Role of Expert Knowledge in Uncertainty Quantification

**(Are We Adding More Uncertainty  
or More Understanding?)**

Jane M. Booker, ESA-WR

Mark C. Anderson, DX-5

Mary A. Meyer, D-1

# Expert Knowledge

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**Expert Knowledge:** what is known by qualified individuals, responding to complex, difficult (technical) questions, obtained through formal expert elicitation.

- A snapshot of the expert's state of knowledge at the time.
- Expressed in qualitative and quantitative form.

# Expert Knowledge = Expertise + Expert Judgment

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## Structure (*Expertise*)

- Define the problem
- Organize and represent the problem solving knowledge, the information flow
- Identify the relevant data and information (e.g., models, experimental results, numerical methods. . .)
- Identify **uncertainties** and determine how these are to be represented

## Contents (*Judgment*)

- Provide quantitative and qualitative estimates and uncertainties, and the heuristics, assumptions and information used to arrive at answers to technical questions.

# Uses of Expertise & Judgment

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## Expertise:

- Decision about what variables enter into a statistical analysis
- Decision about which data sets to include in an analysis
- Assumptions used in selecting a model or method
- Decision concerning which forms of uncertainty are appropriate to use (e.g., probability distributions)
- Description of experts' thinking and information sources in arriving at any of the above responses

## Expert Judgment:

- Estimation of an occurrence of an event
- Estimation of the uncertainty of parameter
- Prediction of the performance of some product or process

# Uncertainty Quantification

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**Broad Definition** — the process of characterizing, estimating, propagating, and analyzing various kinds of uncertainty (including variability) for a complex decision problem.

For complex computer and physical models — focuses upon measurement, computational, parameter (including sensitivities of outputs to input values), and modeling uncertainties leading to verification and validation.

# Two Categories of Uncertainty

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- **Aleatory** —

Inherent variation,  
Random,  
Irreducible  
(Includes variability)

- **Epistemic** —

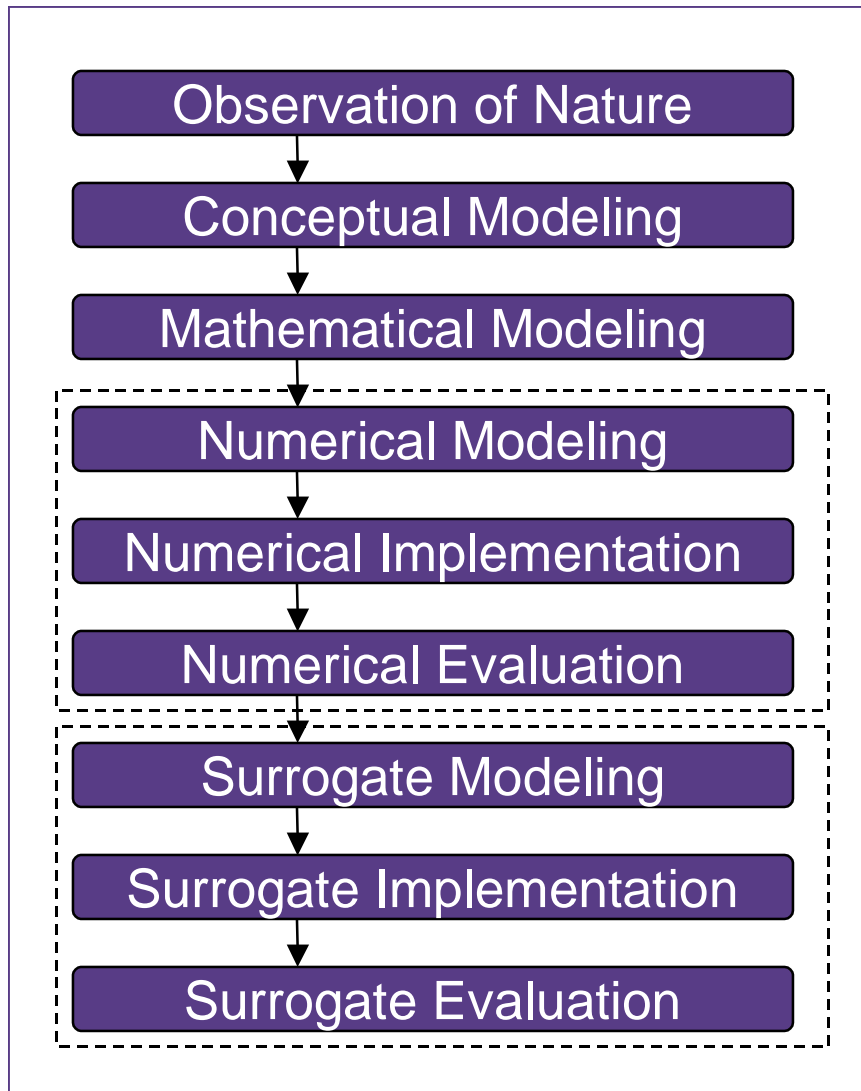
Lack of knowledge,  
Reducible



- **Error** —

numerical,  
discretization,  
mistakes

# The Modeling Process with Uncertainties



## Sources of uncertainty

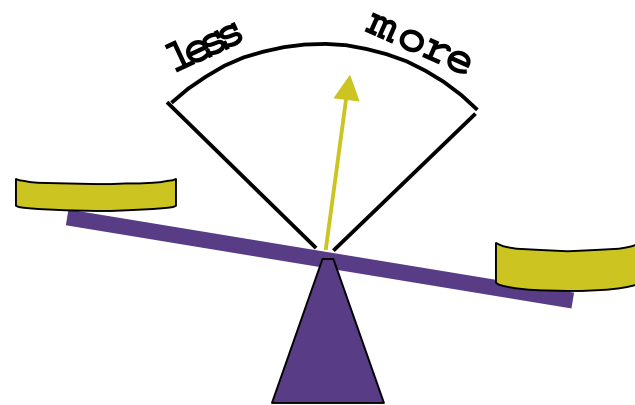
- **Measurements**
  - Noise
  - Resolution
  - Processing
- **Mathematical models**
  - Equations
  - Boundary conditions
  - Initial conditions
  - Inputs
- **Numerical models**
  - Weak formulations
  - Discretizations (mesh, time step)
  - Approximate solution algorithms
  - Truncation and roundoff
- **Surrogate models (statistical)**
  - Approximation error
  - Interpolation error
  - Extrapolation error
- **Model parameters**
- **Scenarios**

# Additional Uncertainty: “Human In The Loop”

## Sources of uncertainty

- Measurements
- Mathematical models
- Numerical models
- Surrogate models (statistical)
- Model parameters
- Scenarios

The expert is making **decisions** about all of these choices and inducing uncertainties in the process.





# Cognitive and Motivational Biases Contribute

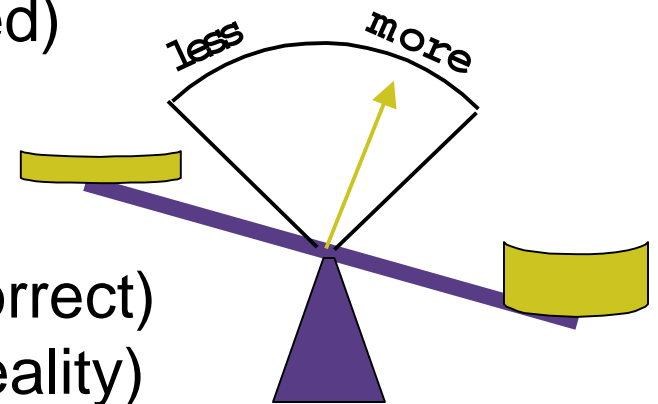
***Bias:*** A skewing from a standard or reference point. Can degrade the quality of the information and contribute to uncertainty.

## Cognitive biases:

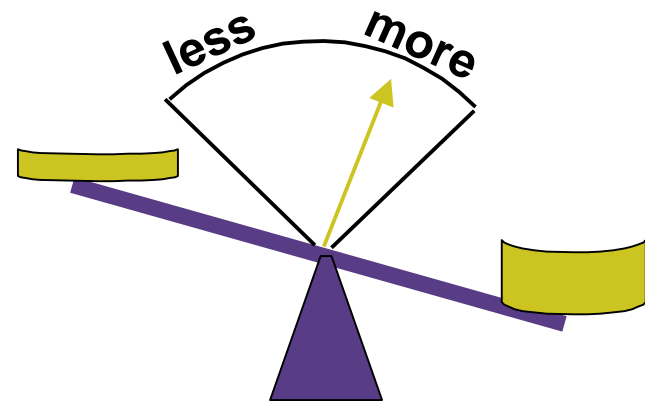
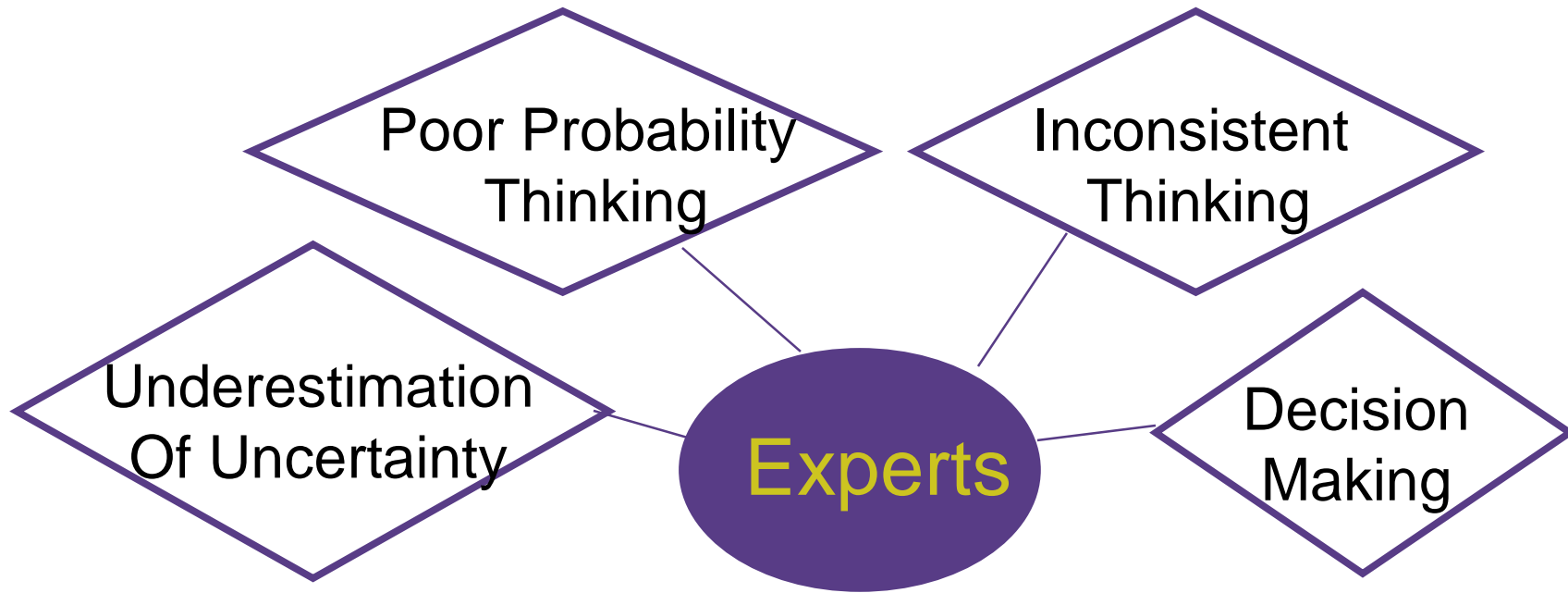
- **Underestimation of uncertainty (false precision)**
- Availability (accounting for rare events)
- Anchoring (cannot move from preconceptions)
- Inconsistency (forgetting what preceded)

## Motivational biases:

- Group think (follow the leader)
- Impression Management (politically correct)
- Wishful thinking (wanting makes it a reality)
- Misrepresentation (bad translation)



# Role of Expert Knowledge in Uncertainty Quantification — Contributions to Uncertainty



# What Tools / Technologies Are Available To Counter These Contributions?

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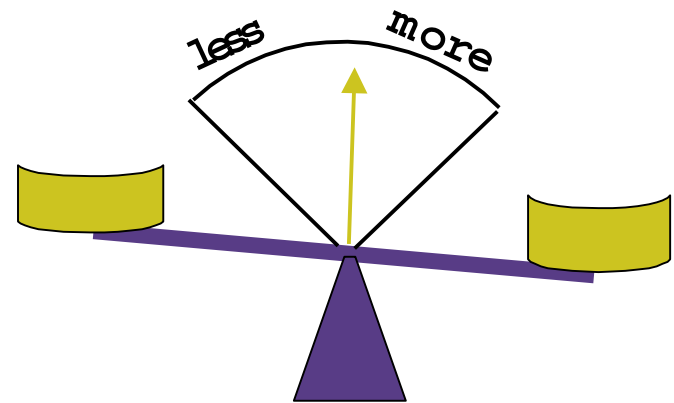
## I. Formal, structured elicitation of expertise and expert judgment

- Draws from cognitive psychology, decision analysis, statistics, sociology, cultural anthropology, and knowledge acquisition.
- Counters common biases arising from human cognition and behavior.
- Adds rigor, defensibility, and increased ability to update the judgments.

# I. Formal, Structured Elicitation of Expertise and Expert Judgment

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- Minimizes biases
- Provides documentation
- Utilizes the way people think, work, and problem solve
- Provides what is necessary for uncertainty quantification:
  - Sources,
  - Quantification,
  - Estimates and Updates,
  - Methods of propagation



## II. Mathematics (Theories) Handling Ignorance, Ambiguity, Vagueness and the Way People Think

- Probability Theory (different interpretations within e.g., Frequentist, Subjective/Bayesian)
- Possibility Theory (crisp or fuzzy set)
- Fuzzy Sets
- Dempster-Schafer (Evidence) Theory
- Choquet Capacities
- Upper and Lower Probabilities
- Convex Sets
- Interval Analysis Theories
- Information Gap Decision Theory (non measure based)

# Mathematical Theories — Frameworks for Expert Thinking

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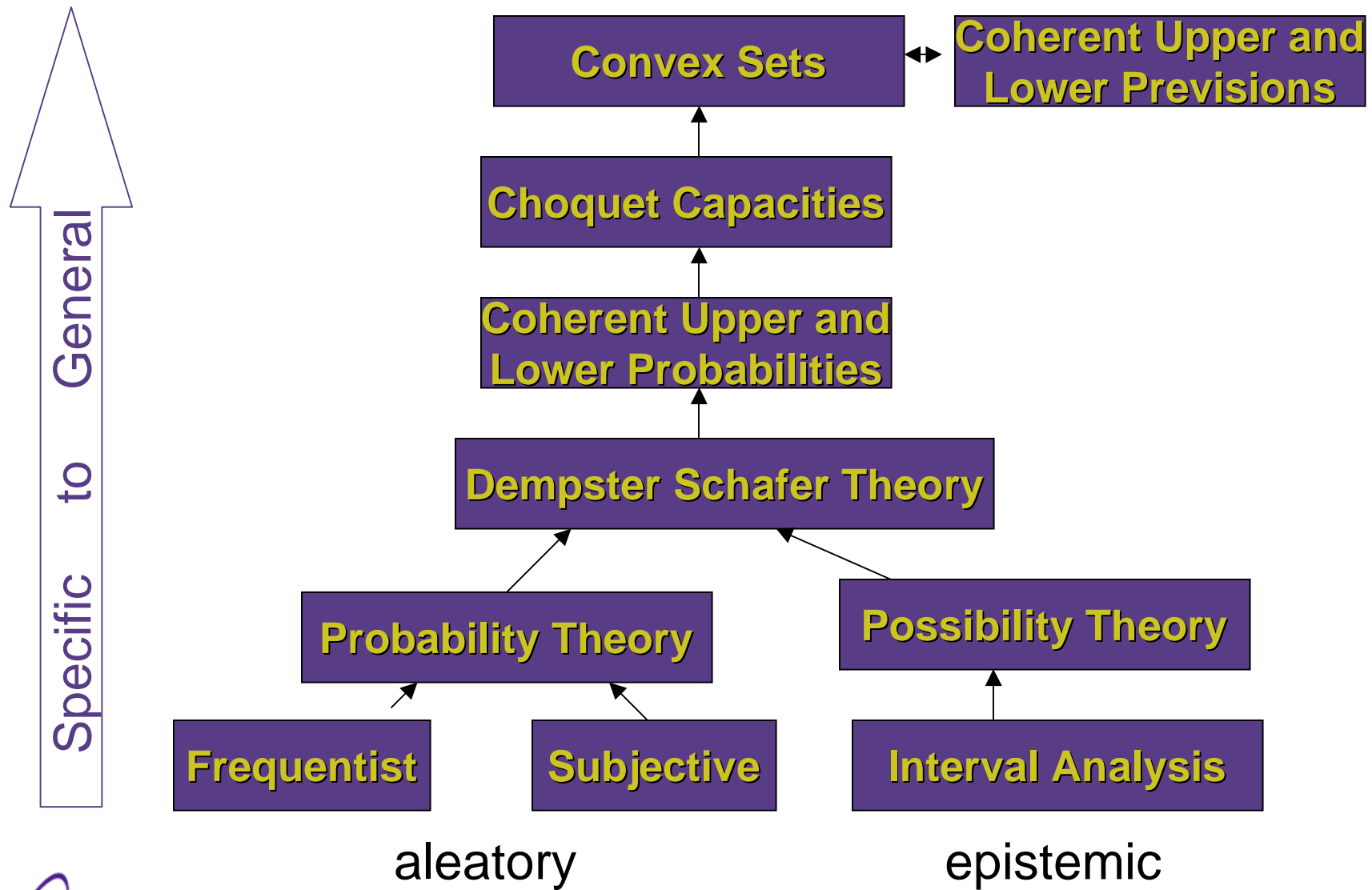
## Characteristics

- Set based (crisp or fuzzy)
- Axiomatic
- Calculus (rules for implementing axioms)
- Consistent / coherence
- Computationally practical (??)
- Measure based (not all!)

Goal: Provide Metrics for Uncertainty

For combining uncertainties there needs to be a bridge between the various theories.

# Hierarchy of Theories for Crisp Sets



# Set Based Theories for Uncertainty

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Non-Measure Based

**Information Gap**

Measure Based

**Fuzzy Sets**

**Crisp Sets**



# Some Measure Theory Approaches

## Probability Theory

Based on single measure function (additivity, monotonic)

$$\Pr: 2^X \rightarrow [0,1]$$

$$\Pr(\emptyset) = 0$$

$$\Pr(X) = 1$$

$$\Pr\left(\bigcup_i A_i\right) = \sum_i \Pr(A_i) - \sum_{j < k} \Pr(A_j \cap A_k) + \dots + (-1)^{n+1} \Pr\left(\bigcap_i A_i\right)$$

$$\Pr\left(\bigcap_i A_i\right) = \sum_i \Pr(A_i) - \sum_{j < k} \Pr(A_j \cup A_k) + \dots + (-1)^{n+1} \Pr\left(\bigcup_i A_i\right)$$

## Dempster-Schafer Theory

Based on two measure functions — belief and plausibility (monotonic & nonadditivity)

$$\text{Bel}: 2^X \rightarrow [0,1] \quad \text{Pl}: 2^X \rightarrow [0,1]$$

$$\text{Bel}(\emptyset) = 0 \quad \text{Pl}(\emptyset) = 0$$

$$\text{Bel}(X) = 1 \quad \text{Pl}(X) = 1$$

$$\text{Bel}\left(\bigcup_i A_i\right) \geq \sum_i \text{Bel}(A_i) - \sum_{j < k} \text{Be}(A_j \cap A_k) + \dots + (-1)^{n+1} \text{Be}\left(\bigcap_i A_i\right)$$

$$\text{Pl}\left(\bigcap_i A_i\right) \leq \sum_i \text{Pl}(A_i) - \sum_{j < k} \text{Pl}(A_j \cup A_k) + \dots + (-1)^{n+1} \text{Pl}\left(\bigcup_i A_i\right)$$

## Possibility Theory

Based on two measure functions — possibility & necessity (monotonic & nonadditivity)

$$\text{Pos}: 2^X \rightarrow [0,1] \quad \text{Nec}: 2^X \rightarrow [0,1]$$

$$\text{Pos}(\emptyset) = 0 \quad \text{Nec}(\emptyset) = 0$$

$$\text{Pos}(X) = 1 \quad \text{Nec}(X) = 1$$

$$\text{Pos}\left(\bigcup_i A_i\right) = \sup_i \text{Pos}(A_i)$$

$$\text{Nec}\left(\bigcap_i A_i\right) = \inf_i \text{Nec}(A_i)$$

# Potential Uncertainty Metrics

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- Hartley measure for nonspecificity

$$H(A) = \log_2 |A|, \quad |A| \text{ is cardinality of } A$$

- Generalized Hartley measure for nonspecificity in DST

$$N(m) = \sum_{A \in 2^X} m(A) \log_2 |A|, \quad m: 2^X \rightarrow [0,1] \quad m(\emptyset) = 0, \quad \sum_{A \in 2^X} m(A) = 1$$

- U-uncertainty measure for nonspecificity in possibility theory

$$U(r) = \sum_{i=2}^n (r_i - r_{i+1}) \log_2 i, \quad r(x) = \text{Pos}(\{x\}) \quad r_i \geq r_{i+1} \quad \forall i$$

- Shannon entropy for total uncertainty in probability theory

$$S(p) = - \sum_{x \in X} p(x) \log_2 p(x)$$

- Generalized Shannon entropy for total uncertainty in DST

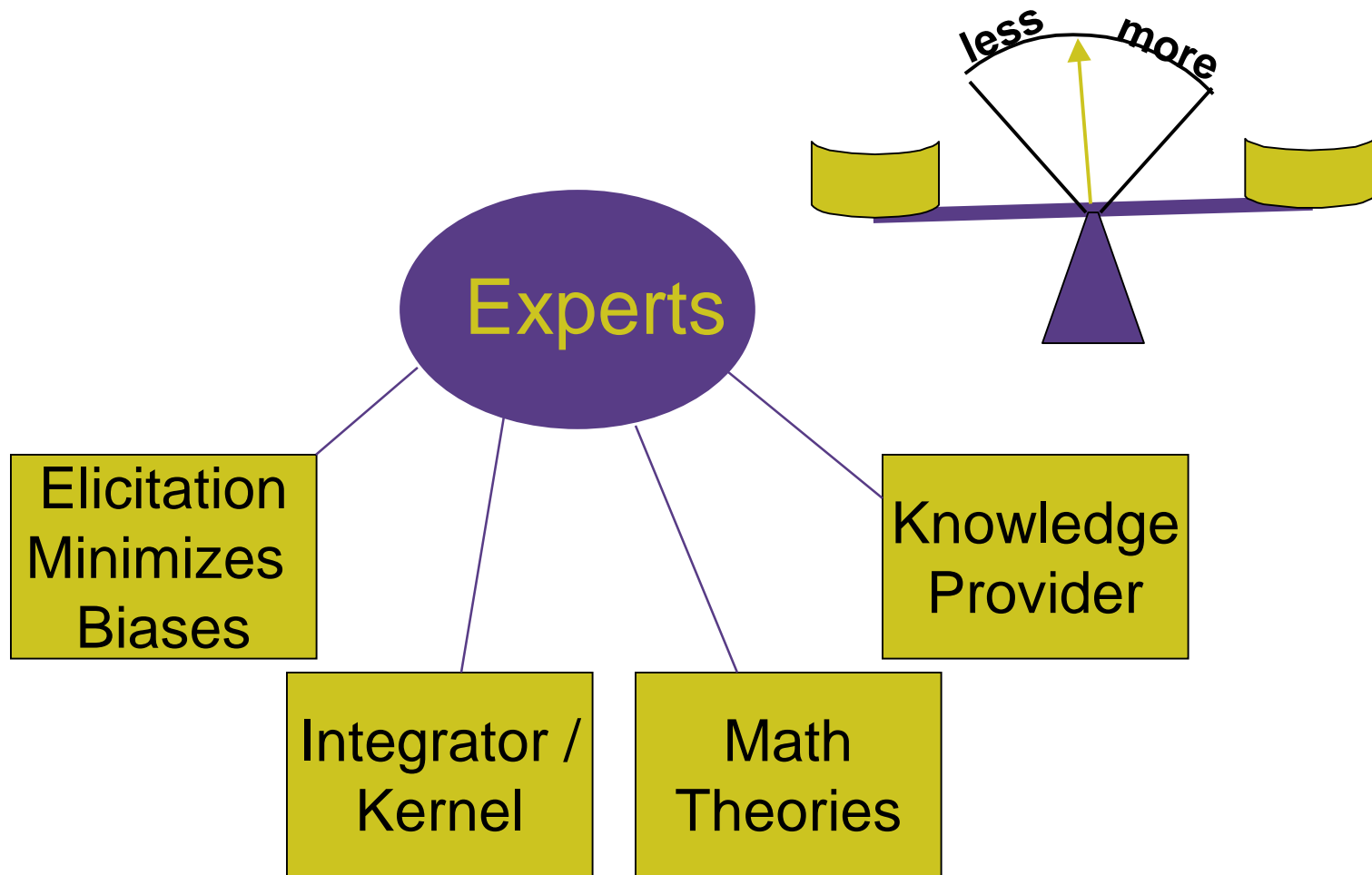
$$AU(\text{Bel}) = \max_{p_x} \left( - \sum_{x \in X} p_x \log_2 p_x \right), \quad \text{Bel}(A) \leq \sum_{x \in A} p_x \quad \forall A \in 2^X$$

- Hamming distance for fuzzy sets

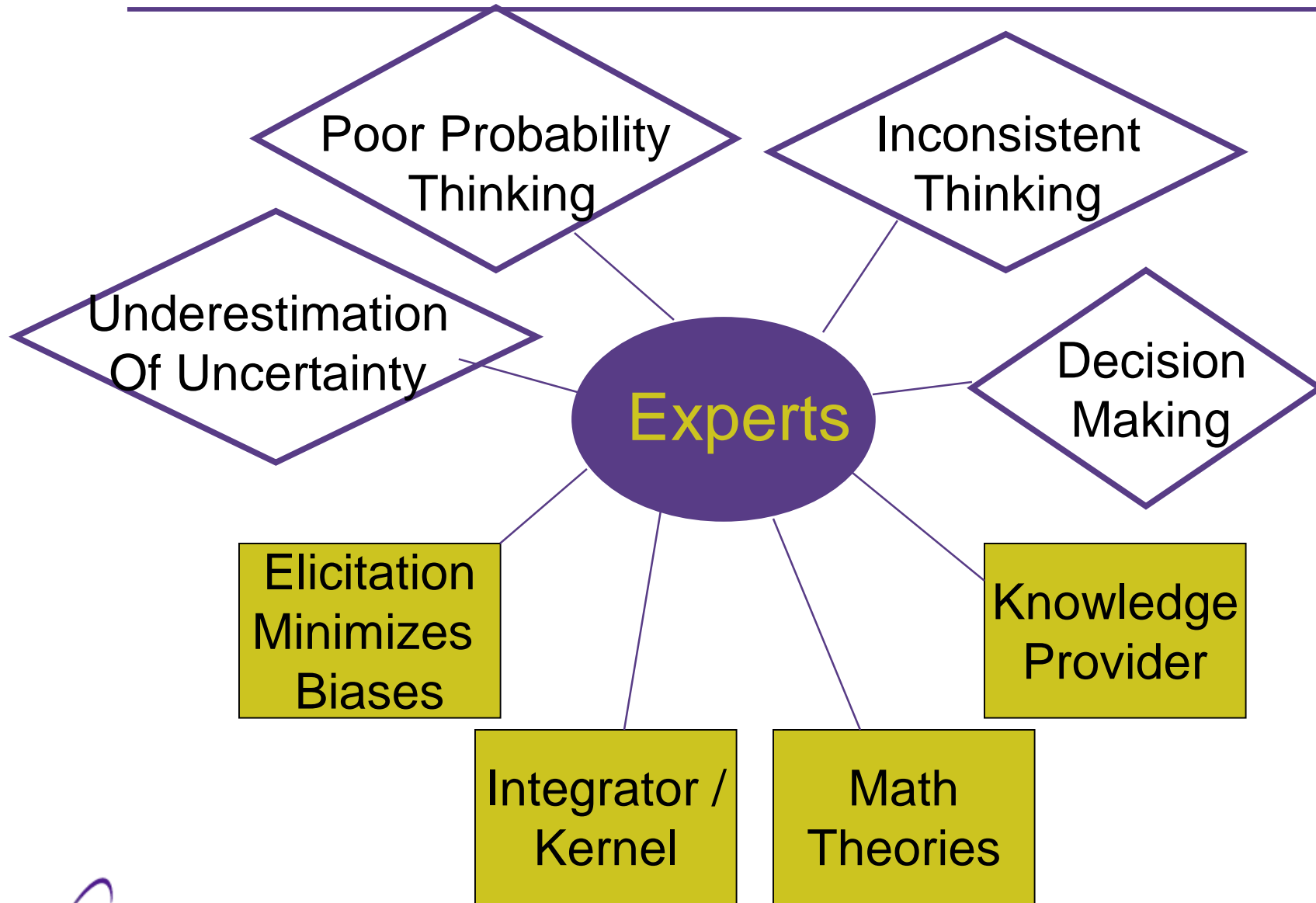
$$f(A) = \sum_{x \in X} [1 - |2A(x) - 1|], \quad A(x) \text{ is membership function}$$

# Role of Expert Knowledge in Uncertainty Quantification — Gains Understanding

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# Role of Expert Knowledge in Uncertainty Quantification — Contributions & Understanding

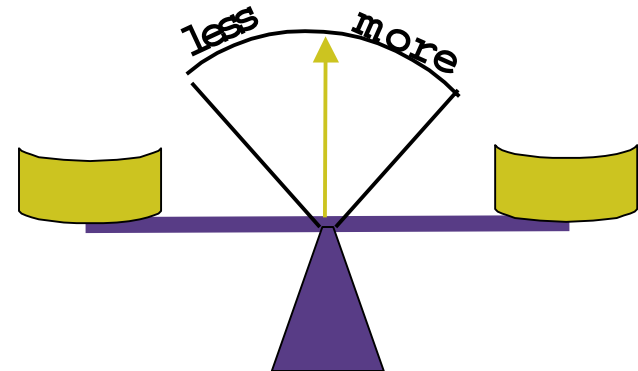


# Role of Expert Knowledge in Uncertainty Quantification

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## Are We Adding More Uncertainty or More Understanding?

A question of balance.



With proper elicitation methods and alternatives probability theory for uncertainties, experts can provide the information, estimation, and integration necessary for understanding uncertainty.