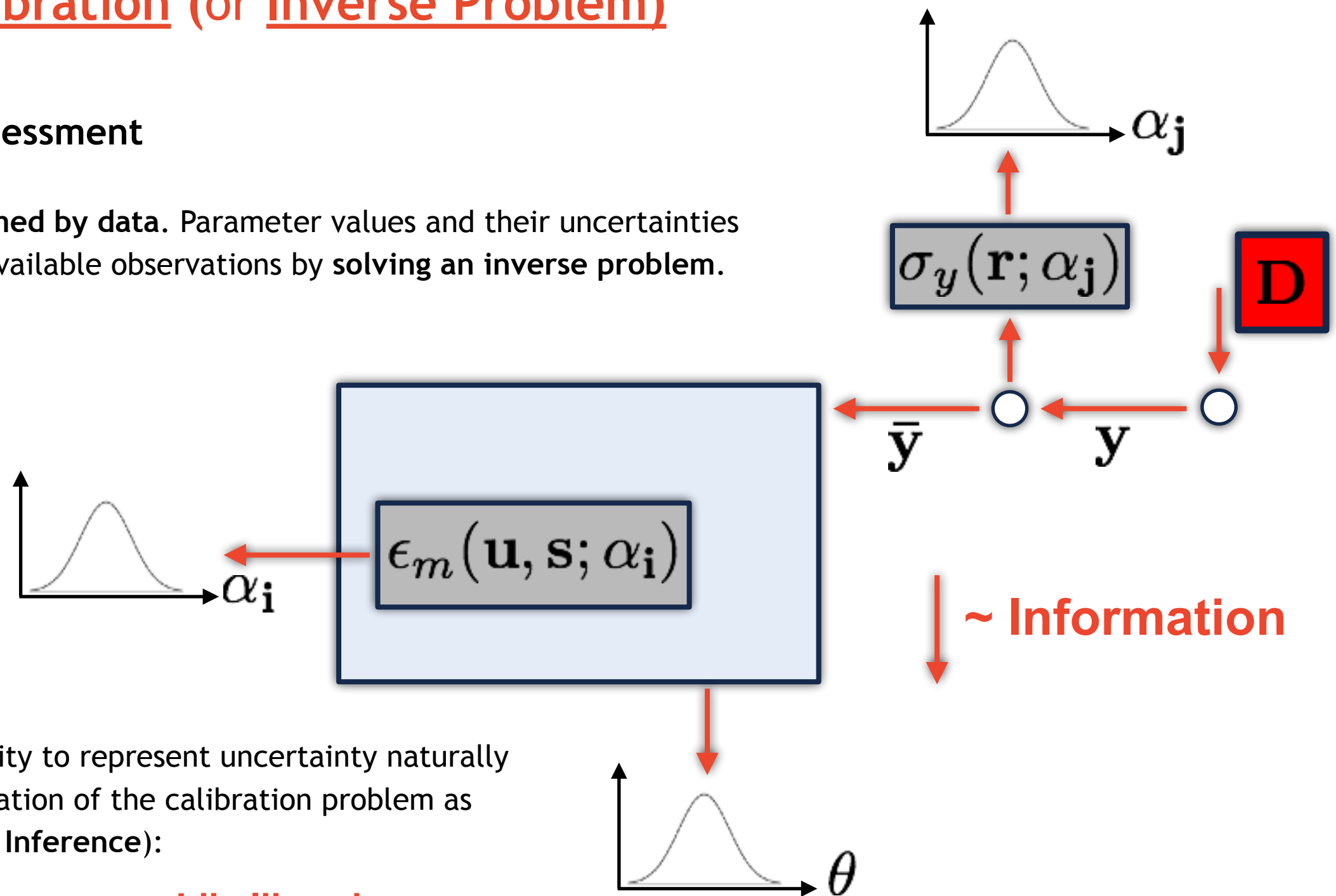


The 3-Steps Reliability Assessment

- **Model Calibration (or Inverse Problem)**

- Validation
- Predictive Assessment

The model is informed by data. Parameter values and their uncertainties are inferred from available observations by solving an inverse problem.



The use of probability to represent uncertainty naturally leads to the formulation of the calibration problem as Bayesian (Bayesian Inference):

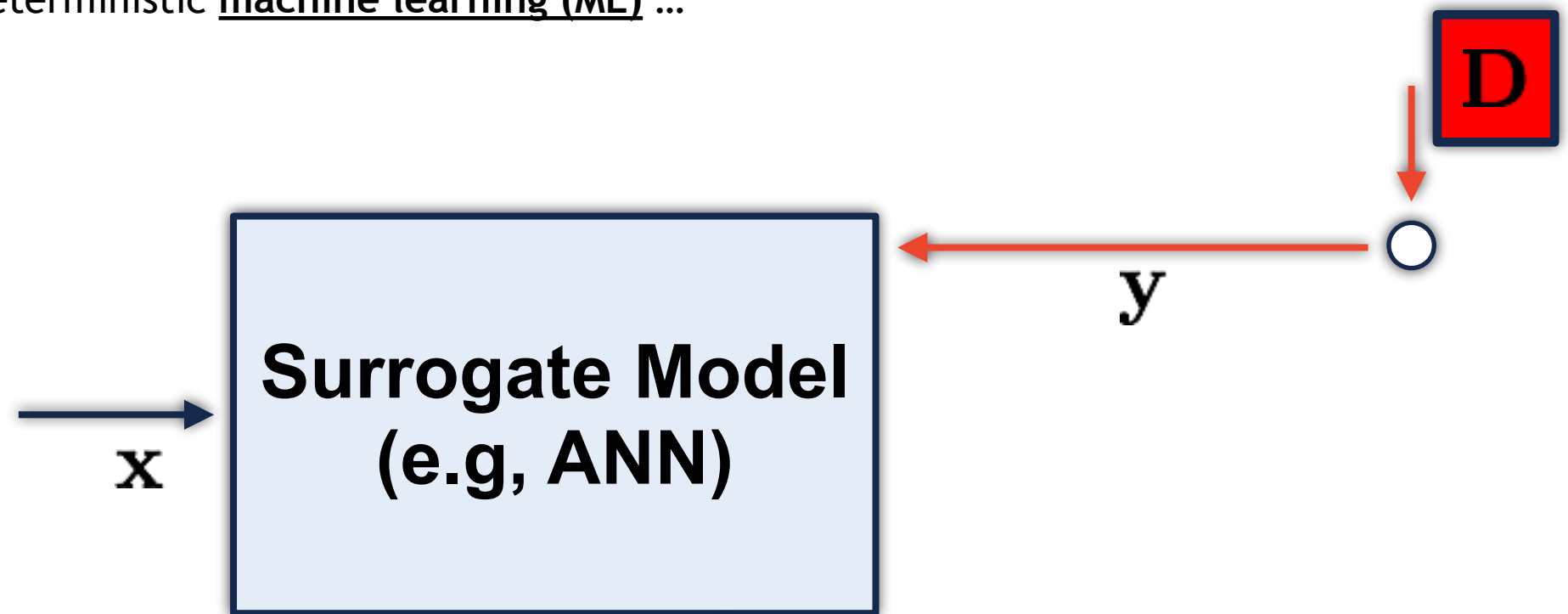
$$\text{Posterior } p(\theta, \alpha | \mathbf{D}, \mathcal{M}) = \frac{\text{Likelihood } \mathcal{L}(\theta, \alpha; \mathbf{D}, \mathcal{M}) \text{ Prior } p(\theta, \alpha | \mathcal{M})}{\int \mathcal{L}(\theta, \alpha; \mathbf{D}, \mathcal{M}) p(\theta, \alpha | \mathcal{M}) d\theta d\alpha}$$

The 3-Steps Reliability Assessment

- **Model Calibration** (or **Inverse Problem**)

- Validation
- Predictive Assessment

Similar to training process in deterministic machine learning (ML) ...



..., but in classic ML:

- The parameters are not treated as random variables;
- The model is deterministic;
- The training happens by means of an optimization algorithm (e.g., ADAMS);

