

Uncertainty Quantification in the Aeroservoelastic Simulation of Wind Turbines

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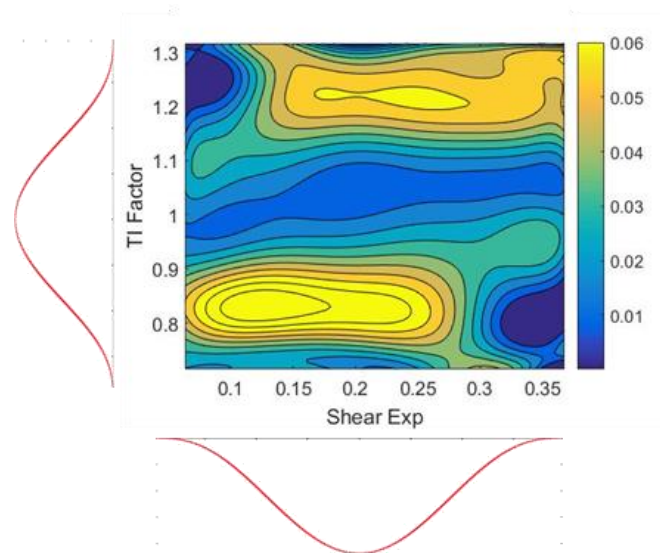
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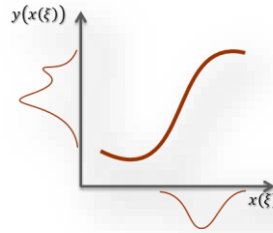
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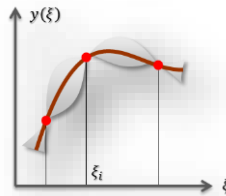


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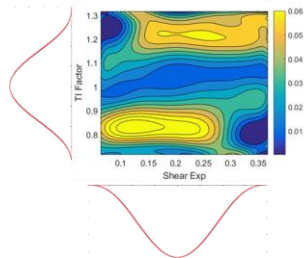
Outline



Motivation: uncertainties in wind turbine simulation and design



Uncertainty Quantification (UQ) methods



Preliminary results



Motivation: Uncertainties in Wind Turbine Modeling

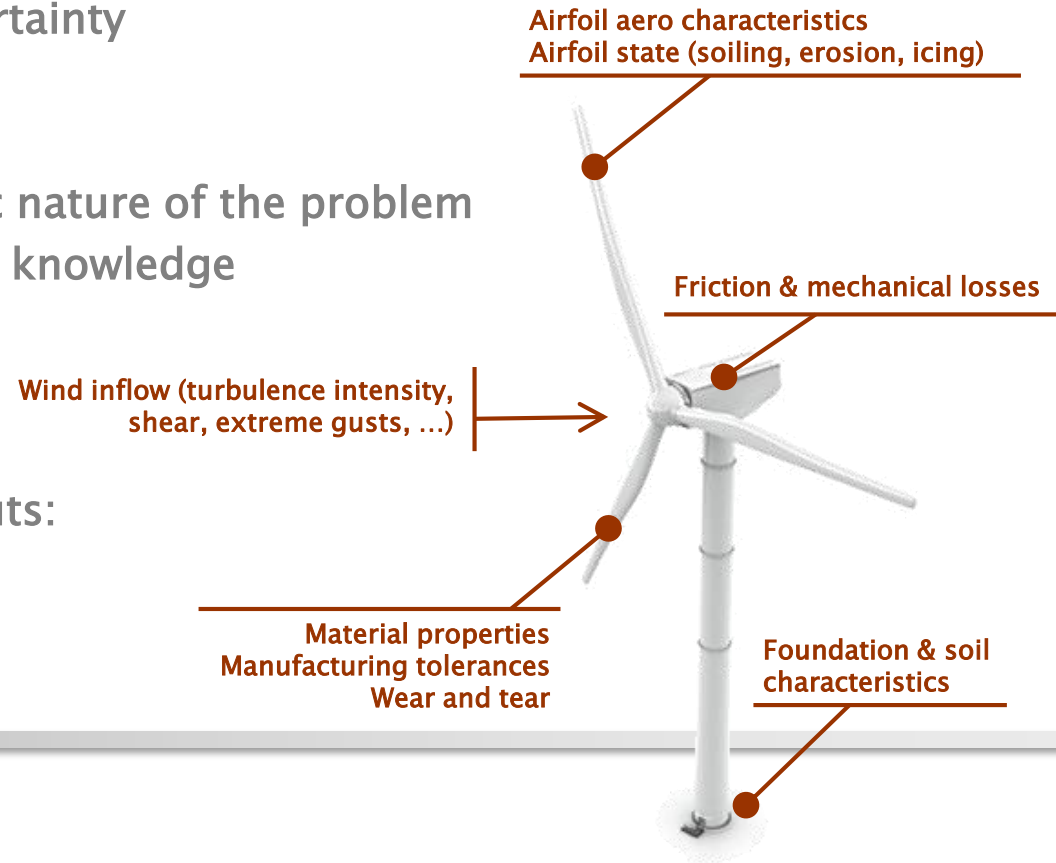
Sources of uncertainty:

- Uncertainty in the model parameters
- Uncertainty in the inputs (e.g., wind conditions)
- Structural uncertainty (model inadequacy)
- Algorithmic uncertainty (numerical errors and approximations)
- Experimental testing uncertainty

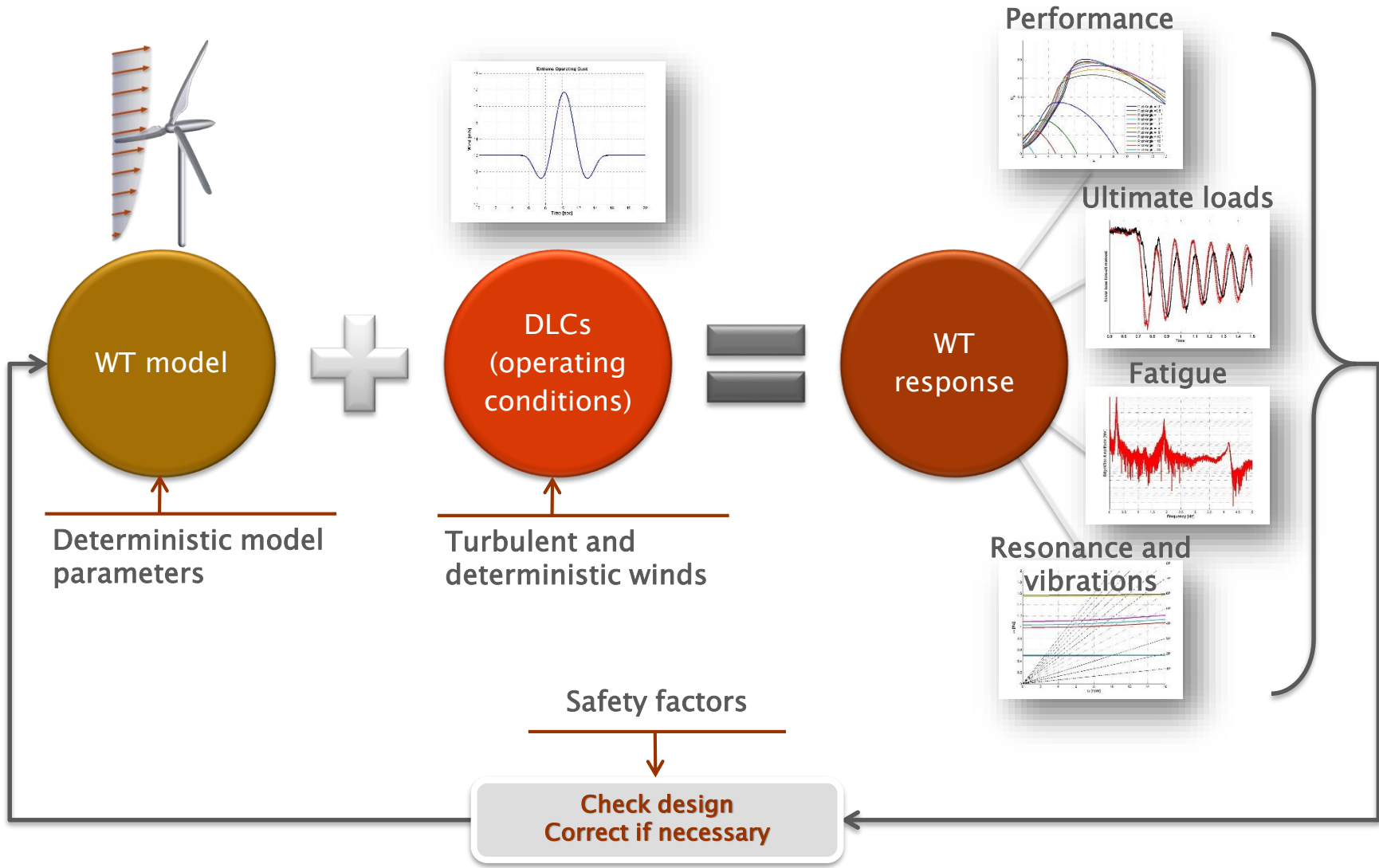
Types of uncertainty:

- Aleatory: due to stochastic nature of the problem
- Epistemic: due to a lack of knowledge

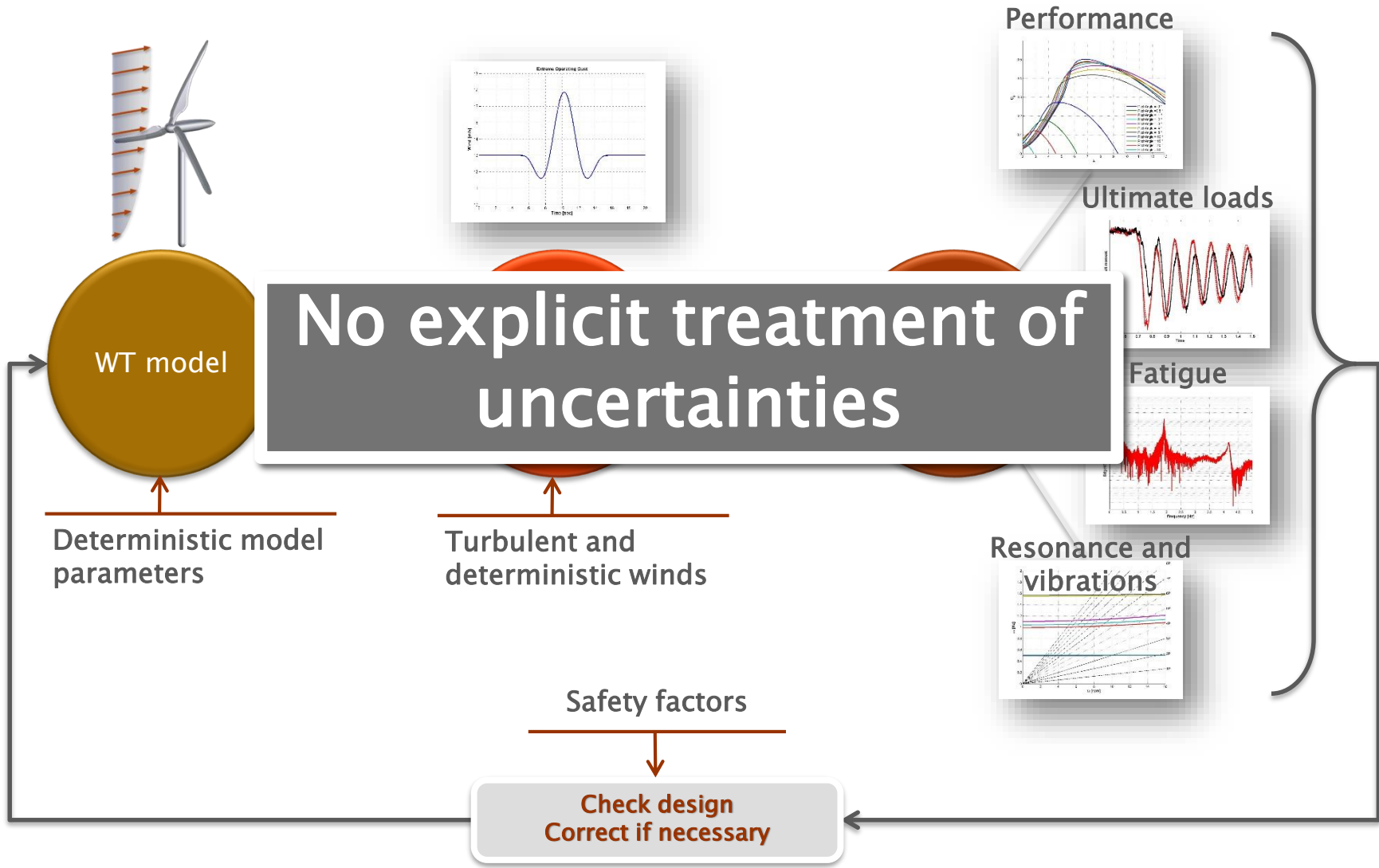
Examples of uncertainties in model parameters and inputs:



Designing According to a Standard



Designing According to a Standard



Long Term Goals

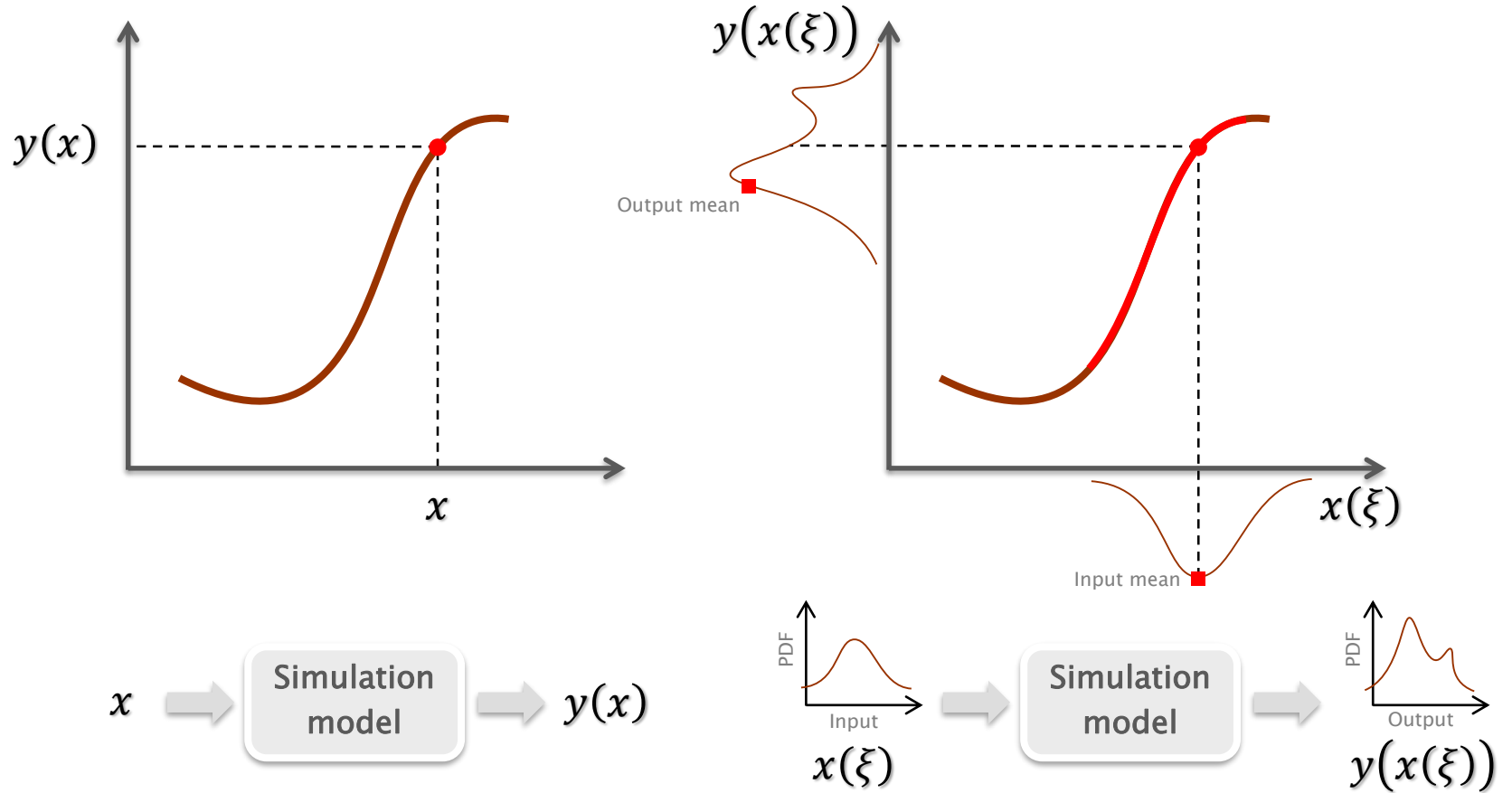
- Improved **understanding** of effects of uncertainties
- Efficient **simulation** in the presence of uncertainties (forward propagation)
- Explicit inclusion of uncertainties in the design process (**robust design**)

This presentation:

- Identification of **best algorithms (applicability to high dimensions, generality, accuracy, few samples)** for forward propagation
- Preliminary assessment of **effects of uncertainties** on design-relevant metrics

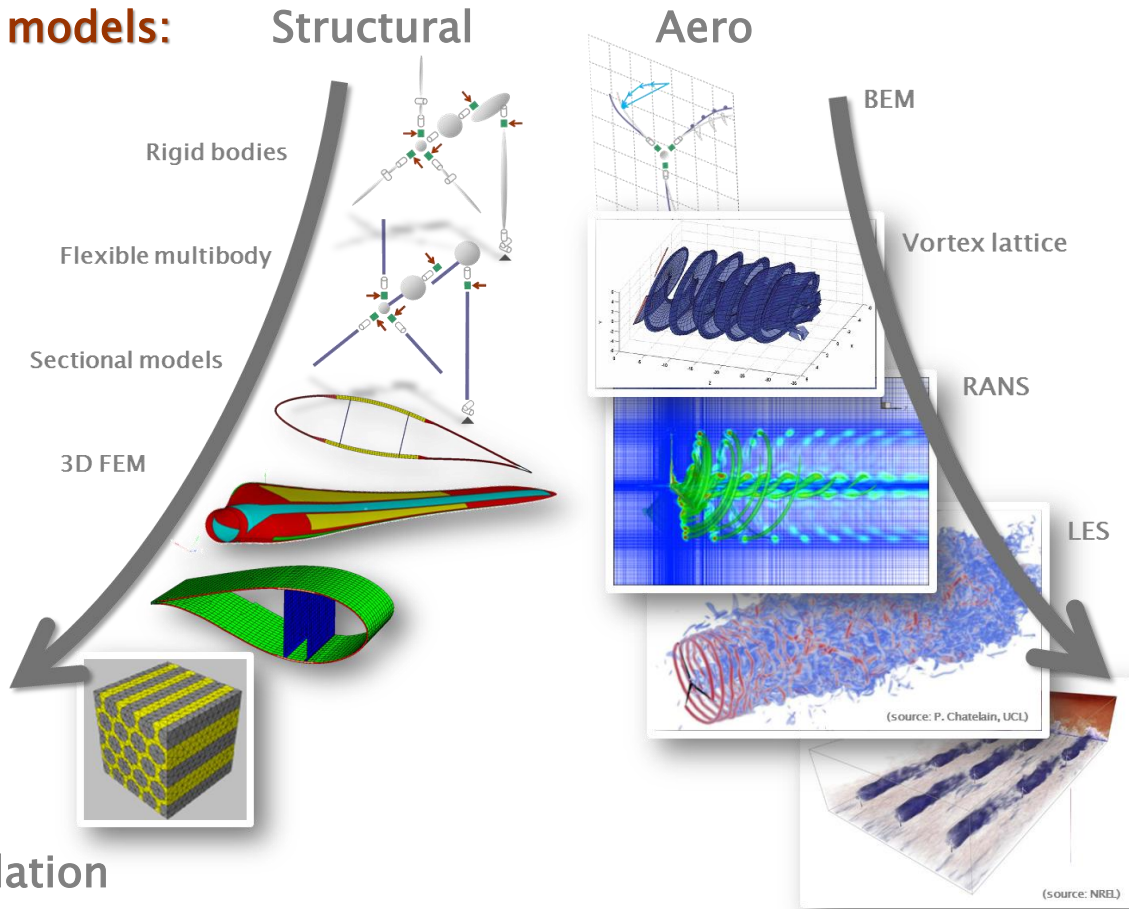


Deterministic vs. UQ-Based Simulation



UQ Forward Propagation Methods

Wind energy simulation models:



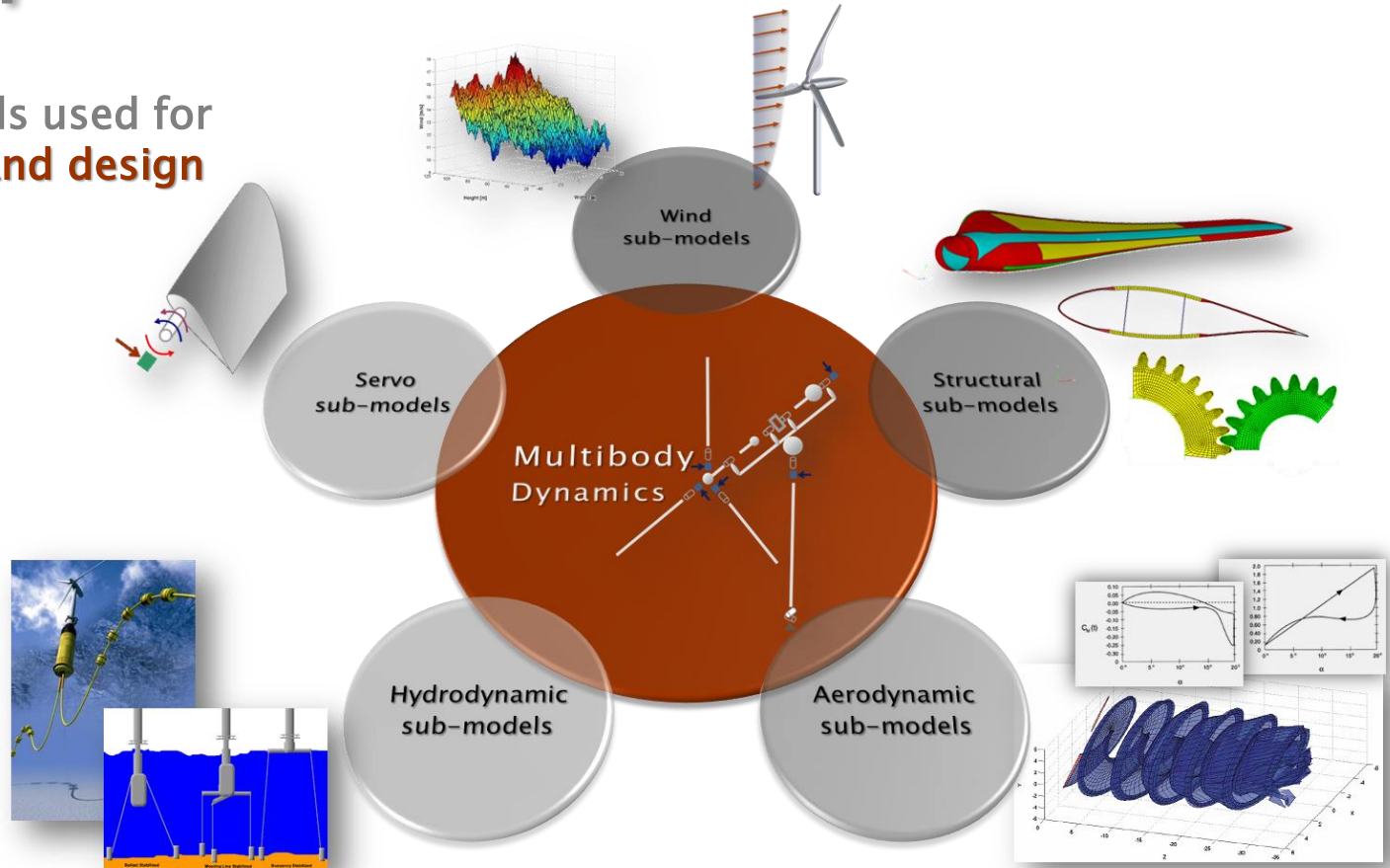
Requirements for UQ:

- Non-intrusive formulation
- Limited computational cost
- Applicable to multiple uncertain inputs (to capture couplings)



Comprehensive Wind Turbine Models

Typical models used for **certification and design**



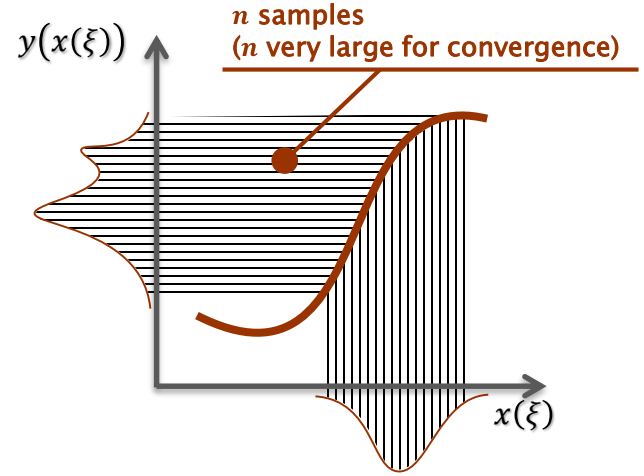
- Medium computational complexity
- Large number of functional evaluations
- Significant computational effort for a complete load assessment



UQ Forward Propagation Methods

UQ methods:

- Simulation-based methods: **Monte Carlo** (MC) (prohibitively expensive, here used for benchmarking -ground truth-)



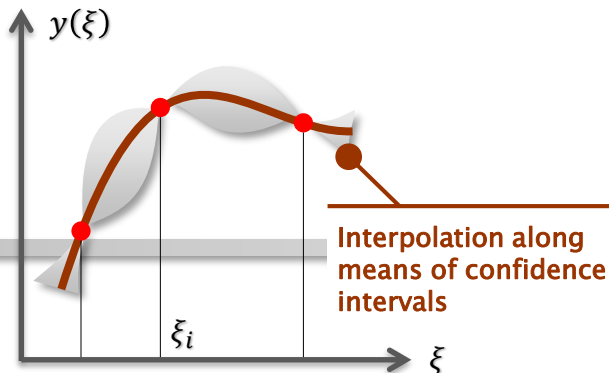
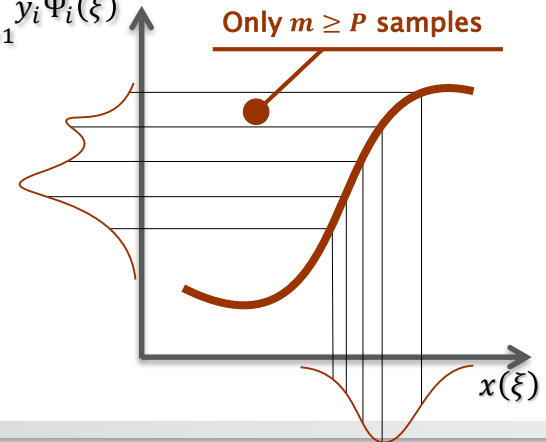
Not suitable

- ~~Local expansion-based methods (only linear models and few uncertain inputs)~~

- Functional expansion-based methods:

- **Non-Intrusive Polynomial Chaos** (NIPC) ▶
- **Kriging** ▶

$$y(\xi) = \sum_{i=1}^P y_i \Psi_i(\xi)$$



Results

Wind turbine:

Data	Value	Data	Value
Wind class	IEC 3A	Rated electrical power	2 MW
Hub height	80 m	Rotor diameter	92 m
Cut-in	4 m/sec	Cut-out	25 m/sec
Solidity	4.6%	Max V_{tip}	72 m/sec



Model: Cp-Lambda aeroservoelastic multibody model (GEB + BEM)

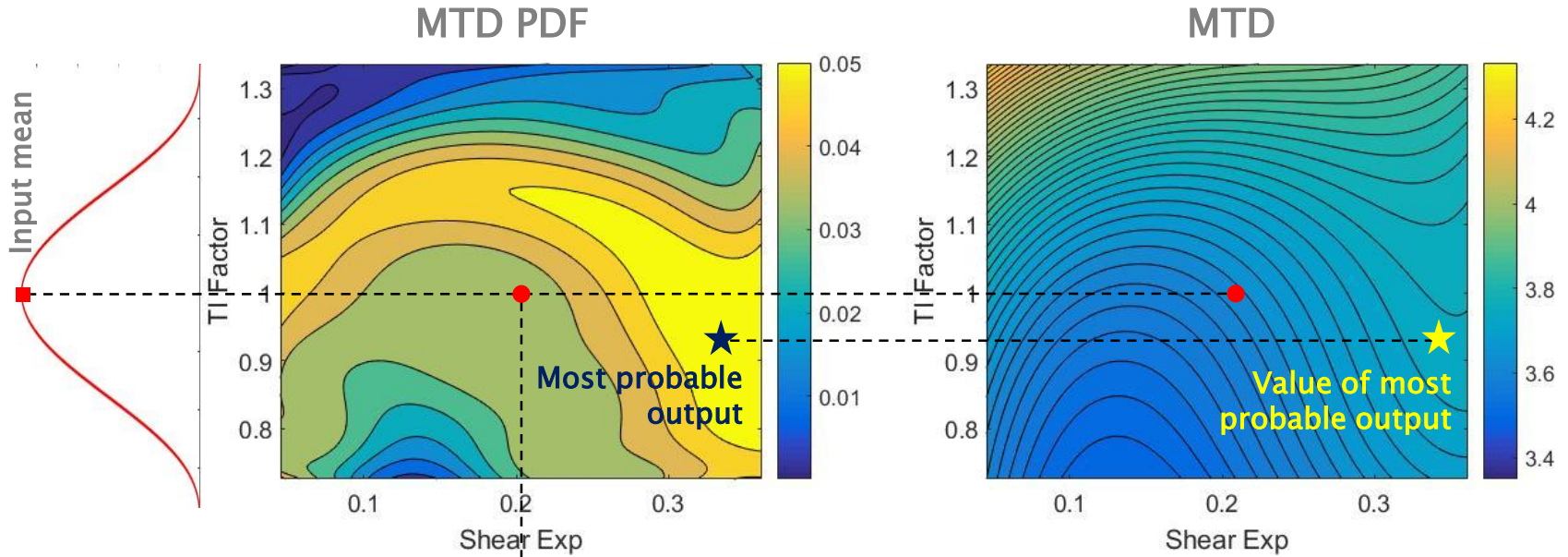
DLCs: IEC 1.1 (Weibull weighting for AEP)

Reference solution: Monte Carlo (>10K samples for convergence of Mean and Std)



Results

Uncertainties: TI & shear; **Method:** NIPC



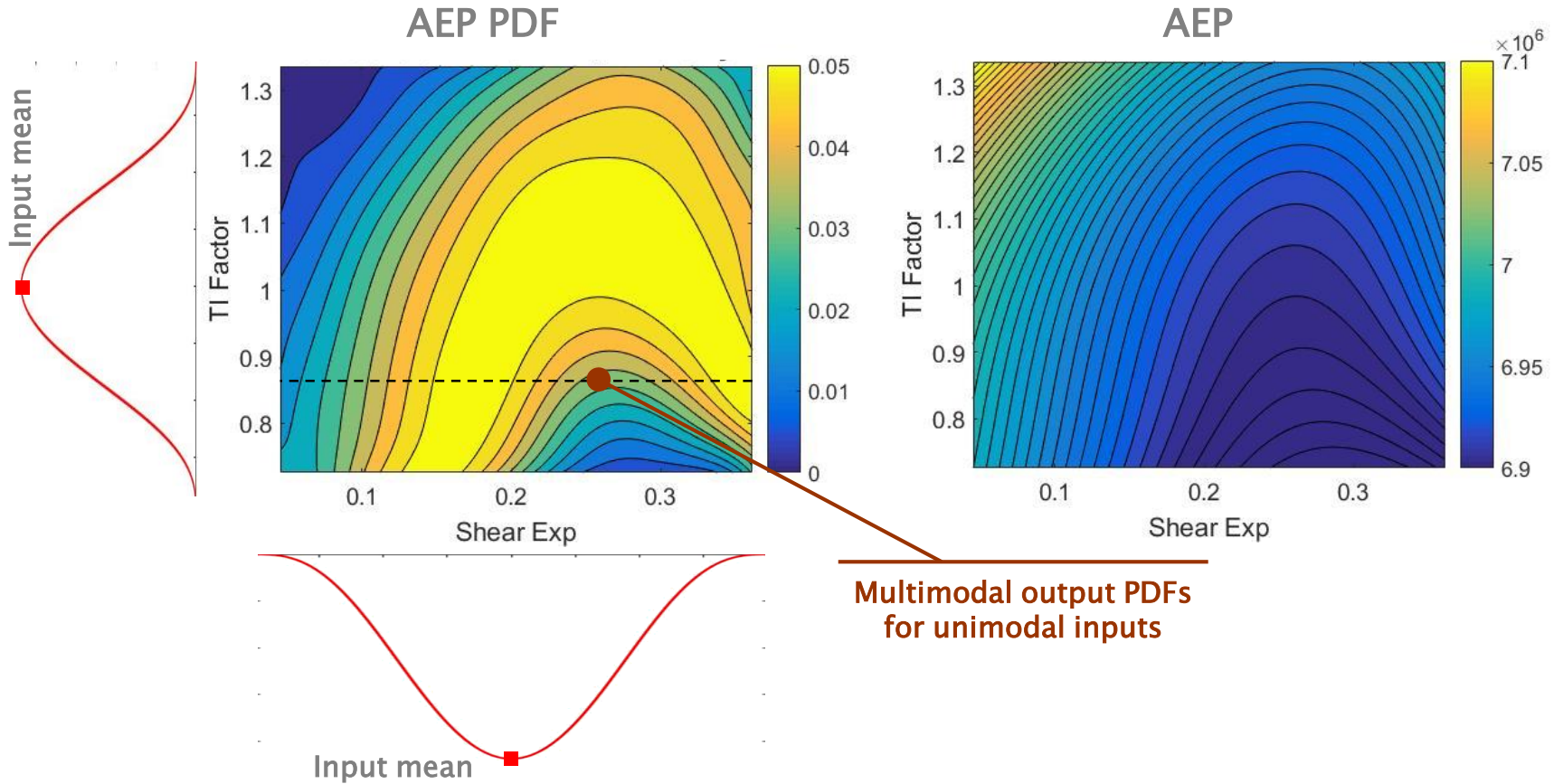
Remark: the most probable MTD is larger than the MTD of the most probable inputs

(MTD: maximum tip deflection)



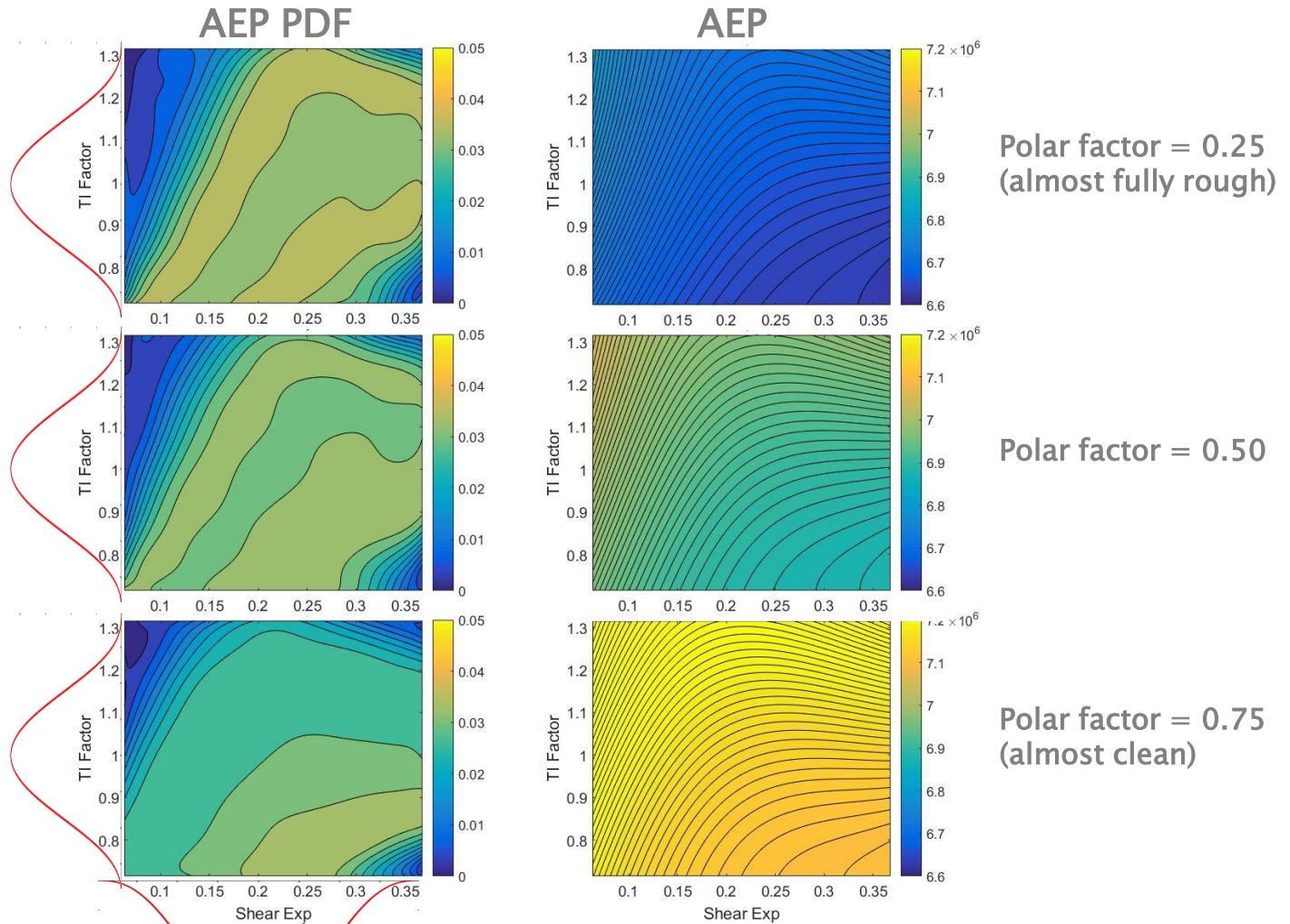
Results

Uncertainties: TI & shear; **Method:** NIPC



Results

Uncertainties: TI, shear & roughness; **Method:** NIPC



Remark: significant couplings among uncertainties



Conclusions

- **Strong non-linearities:** rigorous uncertainty forward propagation methods are necessary
- **Strong couplings** among uncertainties: individual analysis of uncertainties can be misleading
- **Methods:**
 - Both NIPC and Kriging work well at similar costs, small preference for the latter
 - $\mathcal{O}(10^{1\div 2})$ samples, Std convergence slower than Mean

Outlook:

- Use UQ for **extreme load calculation** (avoiding extrapolation)
- **Robust design:**
 - How to include uncertainties in the design process?
 - Would the results be significantly different?

