

CONSTRAINED OPTIMIZATION AND BAYESIAN UPDATING FOR CONDITIONING AND LEARNING

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OBJECTIVES & CHALLENGES

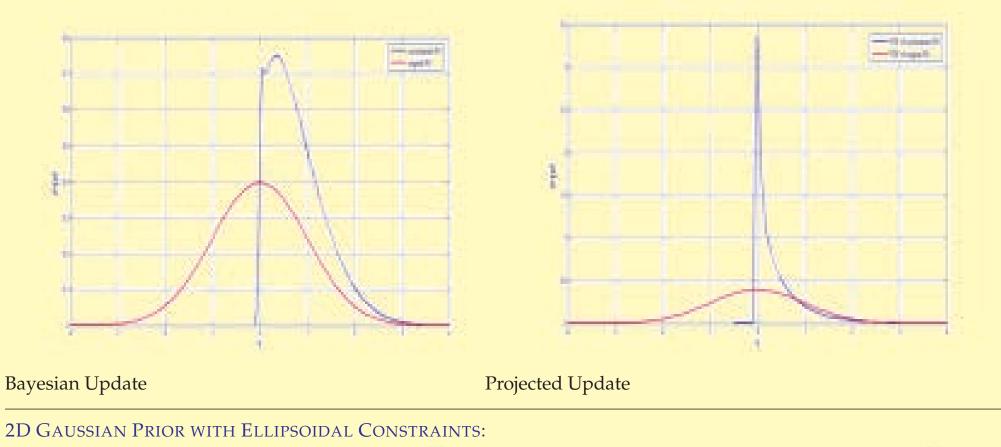
Polynomial Chaos Representations (PCE) are a key tool at the intersection of UQ and CSE. Computing PCE representations, while often computer intensive, yields representations of stochastic processes and random variables that are accurate representations of a functional dependence between random variables as well as carry a convergent approximation of target probability measure.

Traditional probabilistic updating scheme, including Bayesian methods, only require the probabilistic content of these approximation to form the prior and likelihood functions. The functional dependence between input and output, while already computed at great expense, is not leveraged.

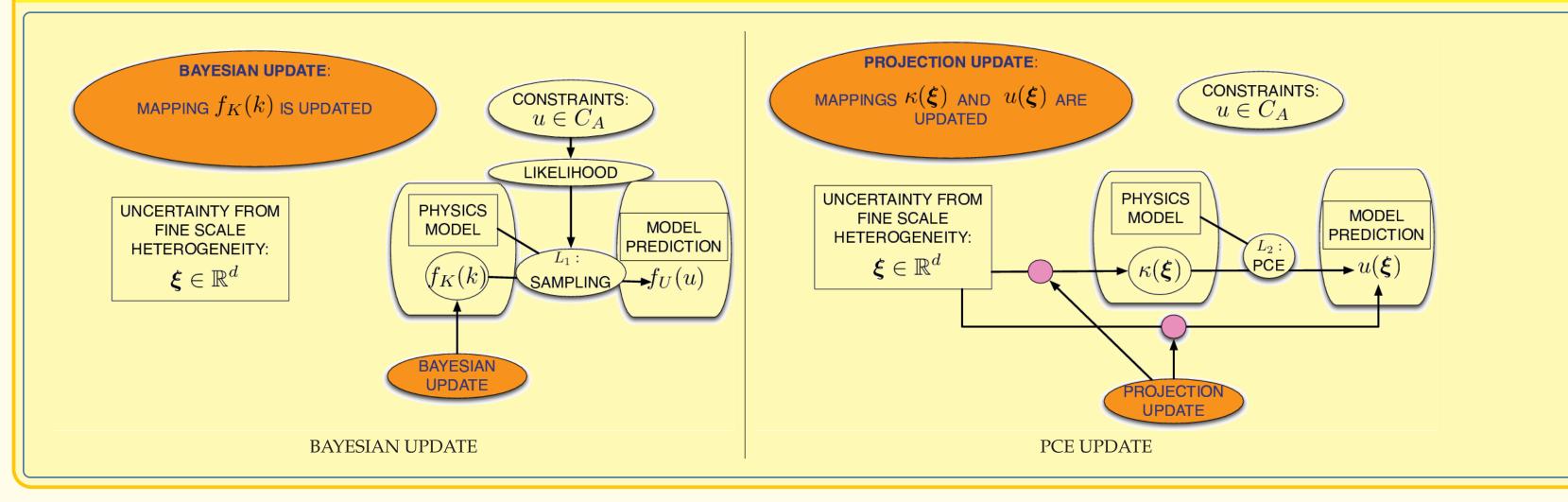
Capitalizing on the PCE construction, updating can be construed as a constrained optimization problem in a Hilbert space already described and constructed as part of the PCE formalism. We demonstrate this projective updating for PCE representations of both output predictions and model parameters.

EXAMPLES

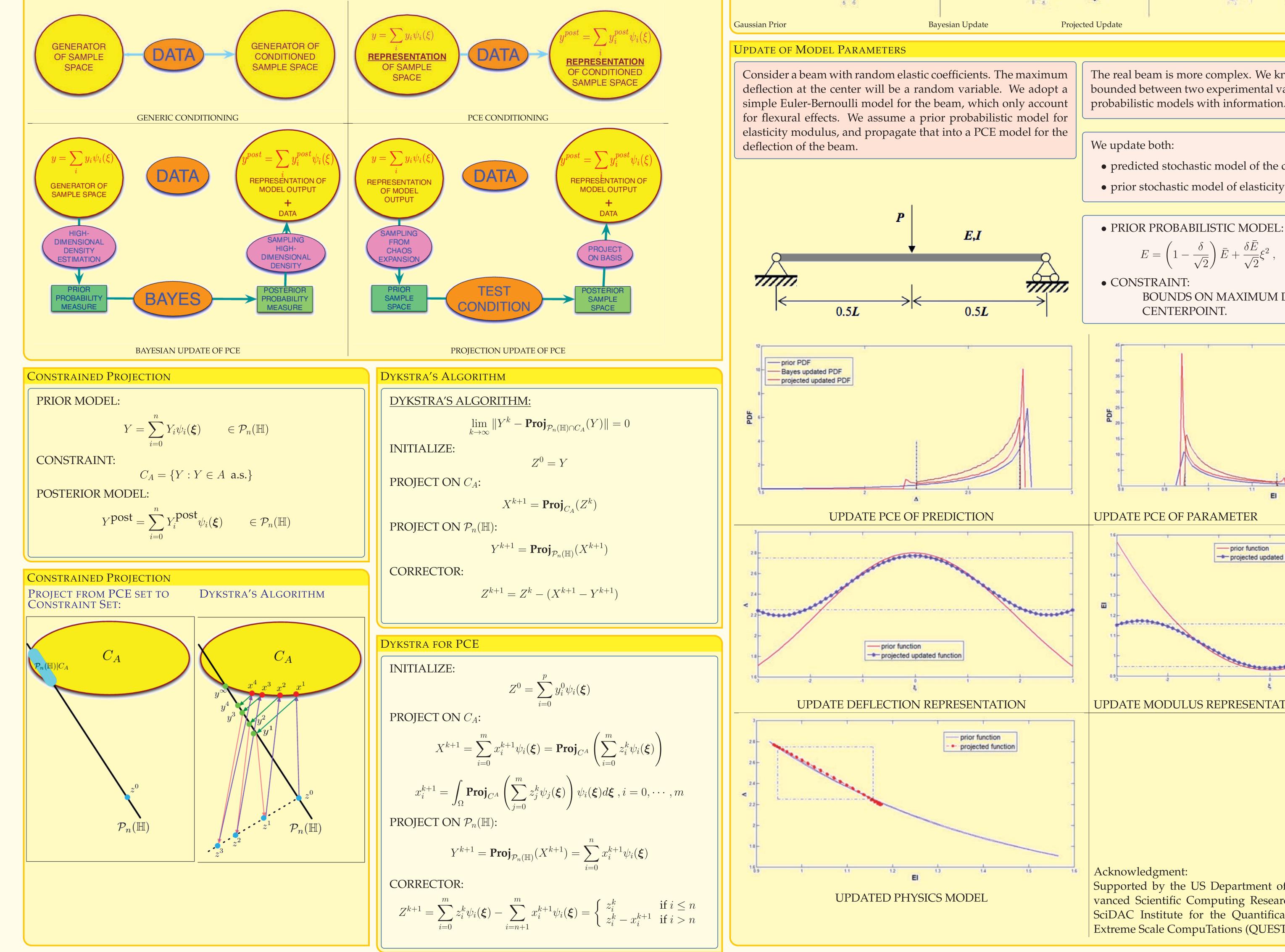
1D GAUSSIAN PRIOR WITH POSITIVE CONSTRAINTS:

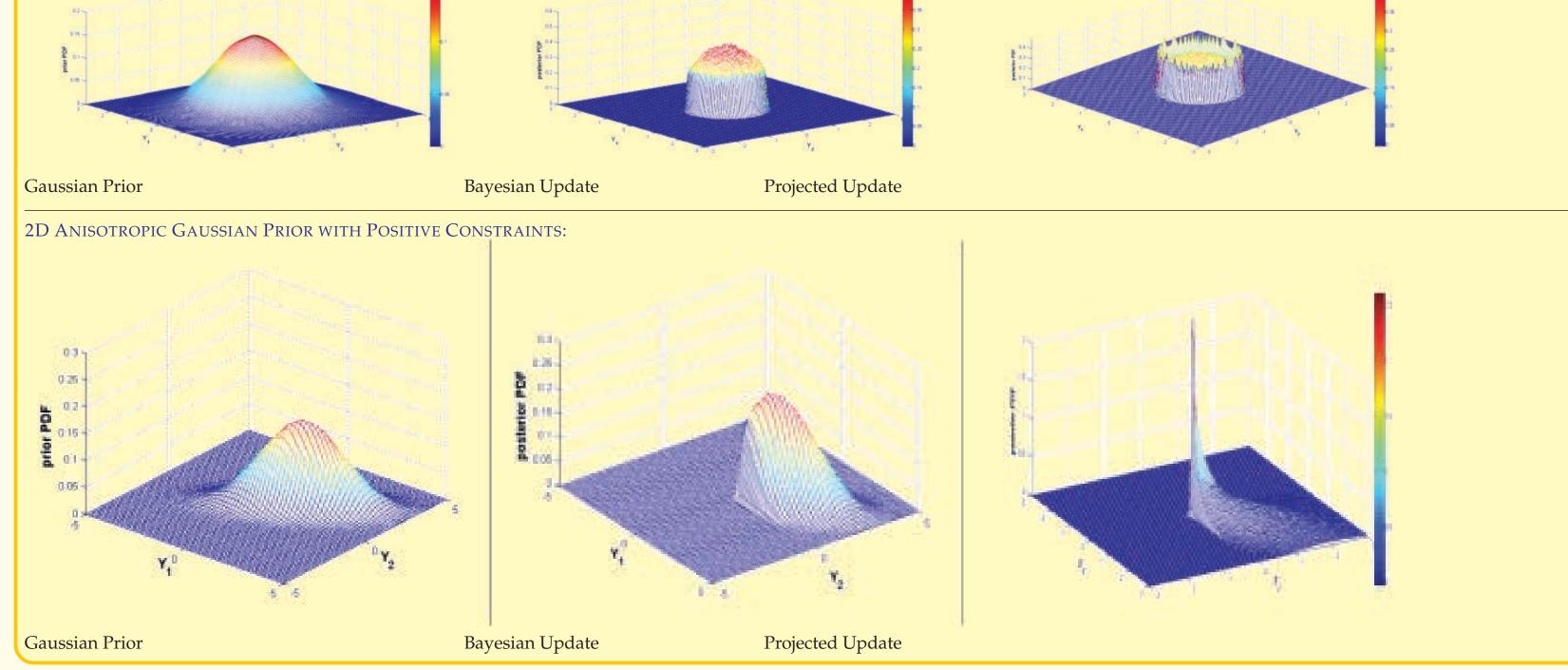


BAYES AND PCE UPDATES



UPDATING POLYNOMIAL CHAOS REPRESENTATIONS



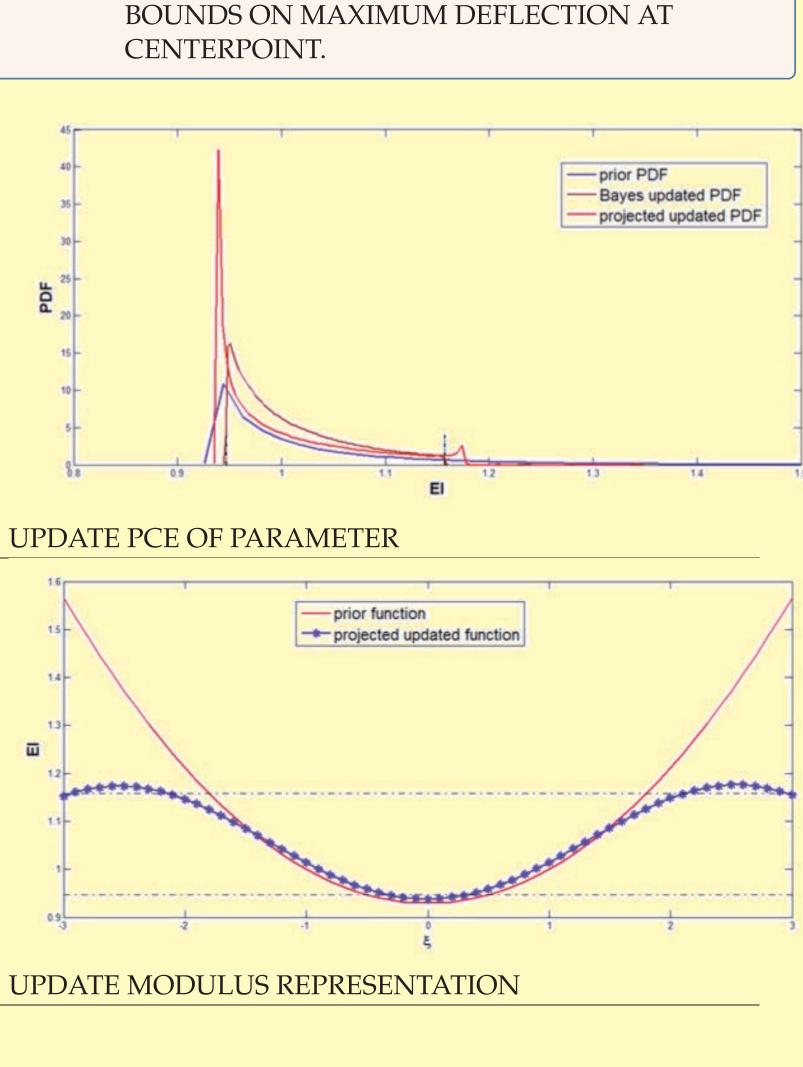


The real beam is more complex. We know its center deflection is bounded between two experimental values. We update the prior probabilistic models with information.

- predicted stochastic model of the deflection.
- prior stochastic model of elasticity parameter.

• PRIOR PROBABILISTIC MODEL:

 $\begin{cases} \xi \sim \mathcal{N}(0,1) \\ \delta = \sigma_E / \bar{E} \end{cases}$



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