

Process of quantifying the effect of uncertainties typically includes:

- **(Global) sensitivity analysis:** identification of input set with greatest influence on output QoIs
- **Uncertainty characterization:** model or infer from observable data; parametric/non-parametric/KDE
- **Uncertainty propagation:** input distributions → output QoI distributions
- **Decision making:** model validation, prediction, design under uncertainty

SNL software tools within QUEST support a range of:

- **UQ studies:** sensitivity analysis, uncertainty propagation, statistical inference
- **Environments:** rapid prototyping in interpreted languages ↔ production computing in compiled languages on parallel platforms
- **Intrusion:** embedded ↔ linked ↔ black box

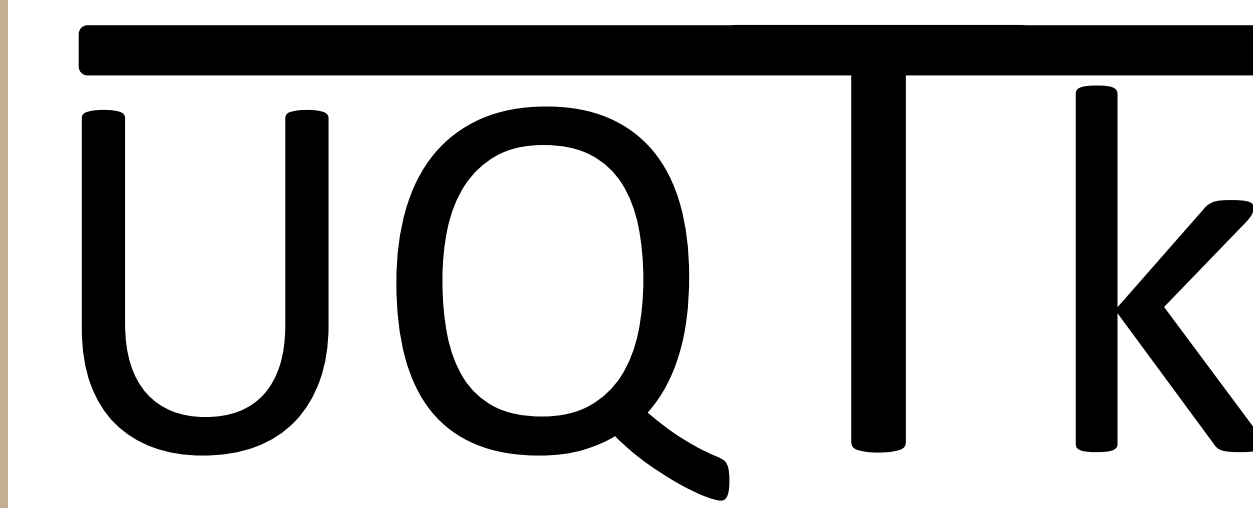
An interoperable set of tools that can be tailored:

- DAKOTA + QUESO/GPMSA with PCE/SC/GP emulators
- Production deployment of stable capabilities in frameworks
- Close collaboration of SAPs with library developers for custom capabilities



DAKOTA (dakota.sandia.gov) is a C++ application that provides a variety of non-intrusive algorithms for design optimization, model calibration, uncertainty quantification, global sensitivity analysis, parameter studies, and solution verification. It can be used as either a stand-alone application or as a set of library services, and supports multiple levels of parallelism for scalability on both capability and capacity HPC resources.

- Contact: dakota-developers@development.sandia.gov



UQTK (www.sandia.gov/UQToolkit) is an LGPL open source library of functions for characterization and propagation of uncertainty in computational models.

- Mainly relies on spectral Polynomial Chaos Expansions (PCEs) for representing random variables and stochastic processes
- Complementary to production tools, UQTK targets:
 - Rapid prototyping
 - Algorithmic research
 - Outreach: Tutorials / Educational
- Contact: Bert Deusschere: bjdebus@sandia.gov

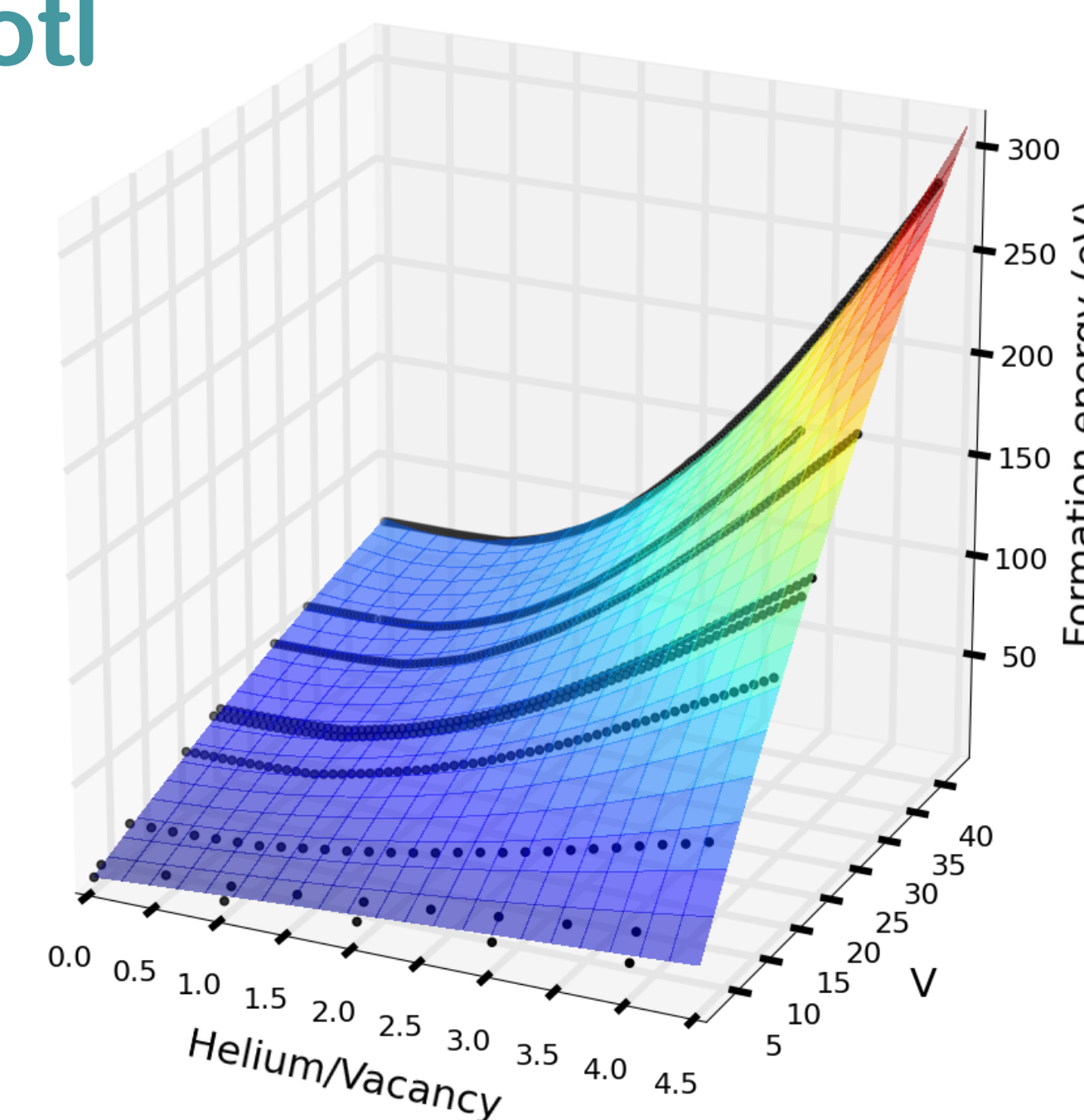
Capabilities:

- Intrusive and non-intrusive (quadrature) approaches for PCE stochastic Galerkin projection
 - Full and sparse quadrature approaches
- Markov Chain Monte Carlo library for Bayesian inference
- Bayesian Compressive Sensing
- Karhunen-Loève expansions
- Sensitivity analysis
- Core libraries in C++
- Examples and postprocessing tools in Python
- Python interface will be released as part of UQTK v3.0, Fall 2015

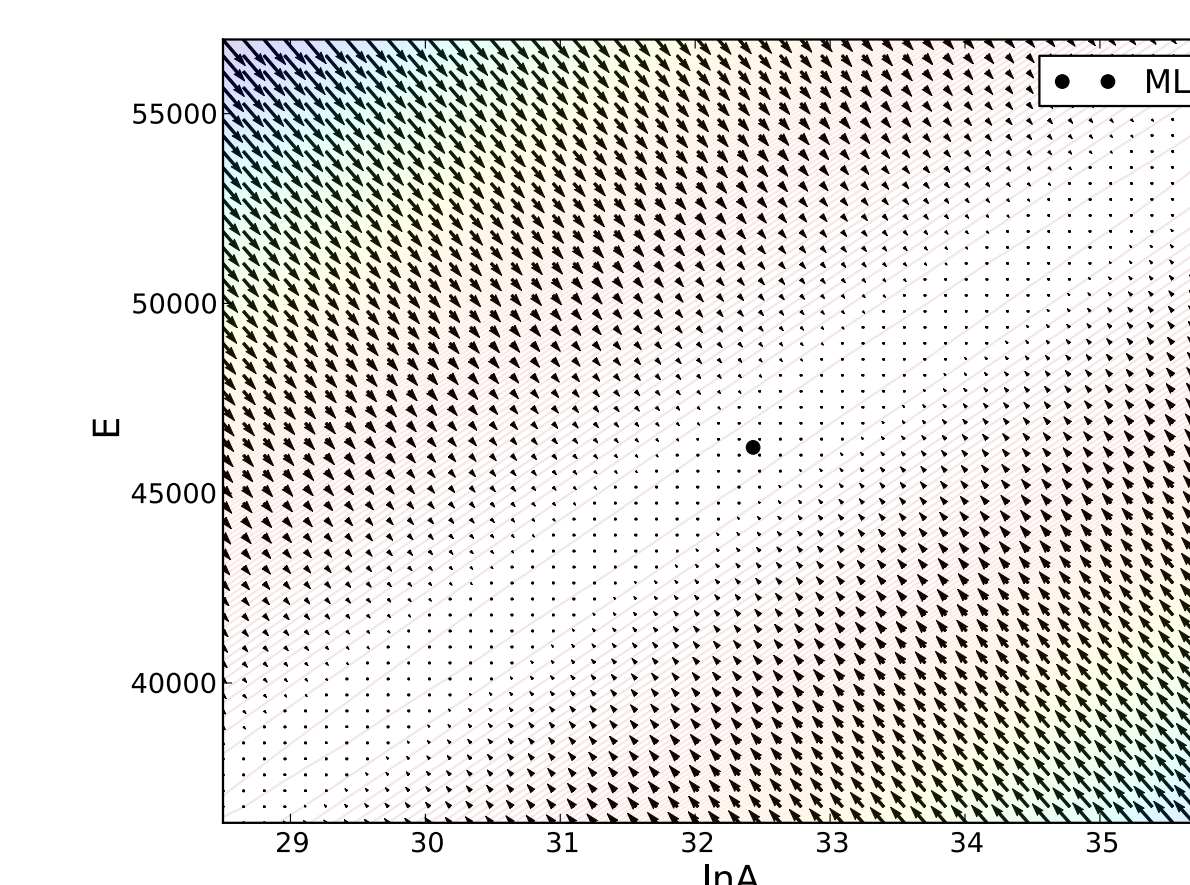
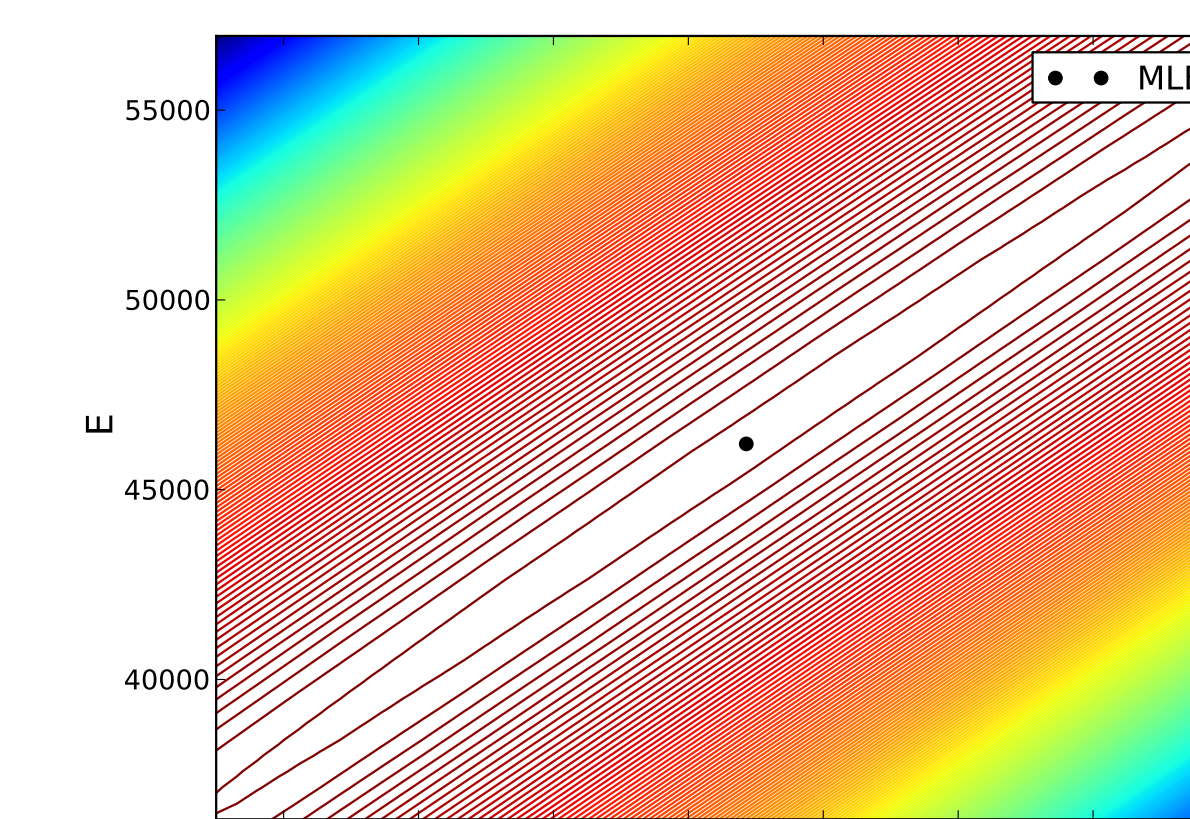
Applications:

- UQTK components can be combined as needed into an end-to-end UQ workflow:
 - Surrogate construction → sensitivity analysis → parameter inference → PCE construction → forward propagation
 - Bayesian compressive sensing used in climate modeling for surrogate construction and dimensionality reduction of land, atmosphere and cloud models (CSSEF, ACME, Multiscale Earth Models, ACES4GCM)
 - UQ workflows set up in multiple SciDAC partnership projects: e.g. UQ in Xolotl (PSI)
- Development of lecture material and hands-on exercises for UQ tutorials
 - Nationally and Internationally

Uncertainty Quantification in Xolotl



Inference of Combustion Model Parameters



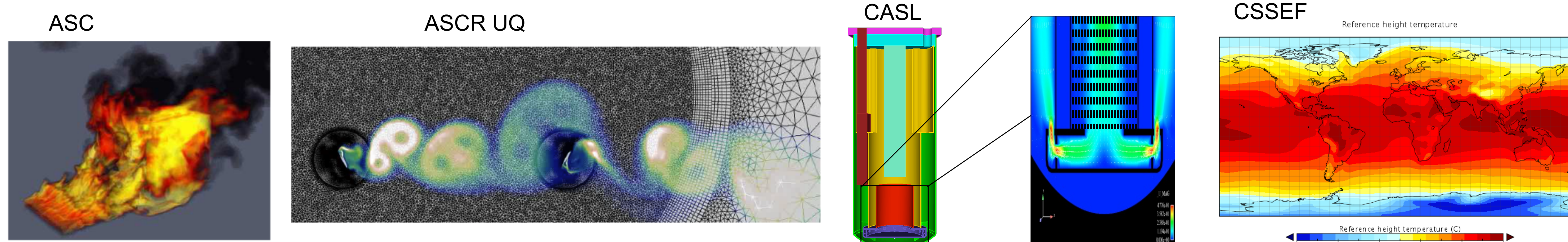
- 2nd order Legendre-Uniform PC surrogate obtained with Bayesian regression from formation energies computed with MD
- Input to Xolotl, which computes cluster dissociation rates in plasma-surface interactions

- 5th order Legendre-Uniform PC surrogate for an ignition time model, as a function of activation energy and pre-exponential (top left)
- Derivative of the surrogate (bottom left)
- Both the surrogate and its derivative obtained with UQTK
- Used in optimization to get better initial guess for MCMC
- Used to accelerate likelihood computation in MCMC

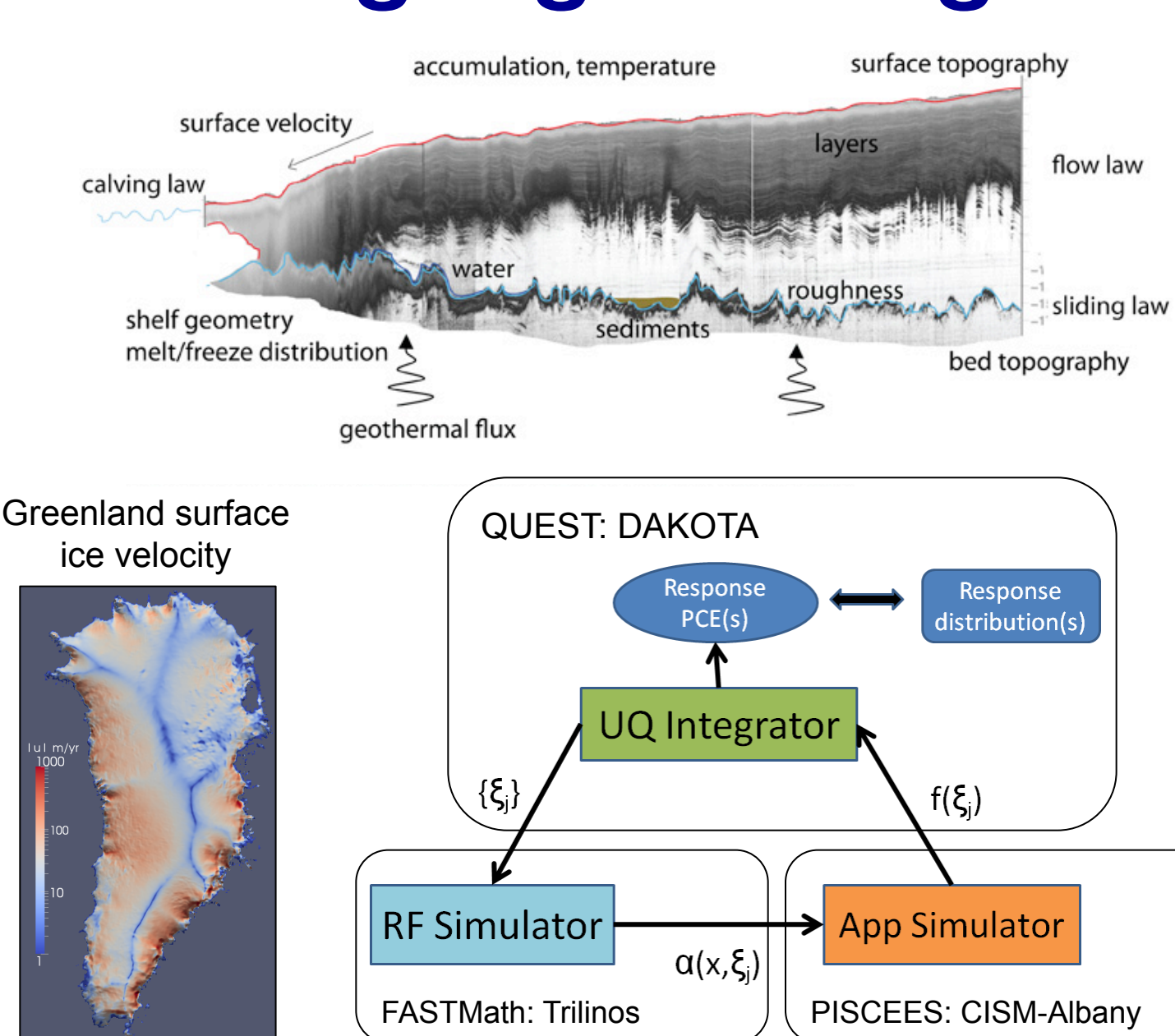
UQ Capabilities in v6.2 (released 5/15/15):

- **Sampling methods**
 - Random: LHS, MC, Incremental
 - Importance: IS, AIS, MMAIS
 - Adaptive: Morse-Smale et al.
- **Reliability methods**
 - Local: MV, AMV, AMV+, AMV²⁺, TANA-3, FORM, SORM
 - Global: EGRA, GPAIS, POF Darts, RKD
- **Stochastic expansion methods**
 - Polynomial chaos: projection, regression (see Algs poster)
 - Stochastic collocation: tensor & sparse; nodal & hierarchical
- **Epistemic methods**
 - Interval estimation: local, global, mixed-integer
 - Dempster-Shafer
- **Bayesian methods**
 - QUESO, GPMSA, DREAM
 - Emulator-based MCMC: PCE, SC, GP (Algs poster)
 - Random field inference (PISCEES at bottom)
- **Meta-iteration and recursion**
 - Mixed aleatory-epistemic UQ
 - Design / calibration under uncertainty

Defense, Science, and Energy Applications



SAP Highlight: Integration of Albany/Dakota/Trilinos for PISCEES



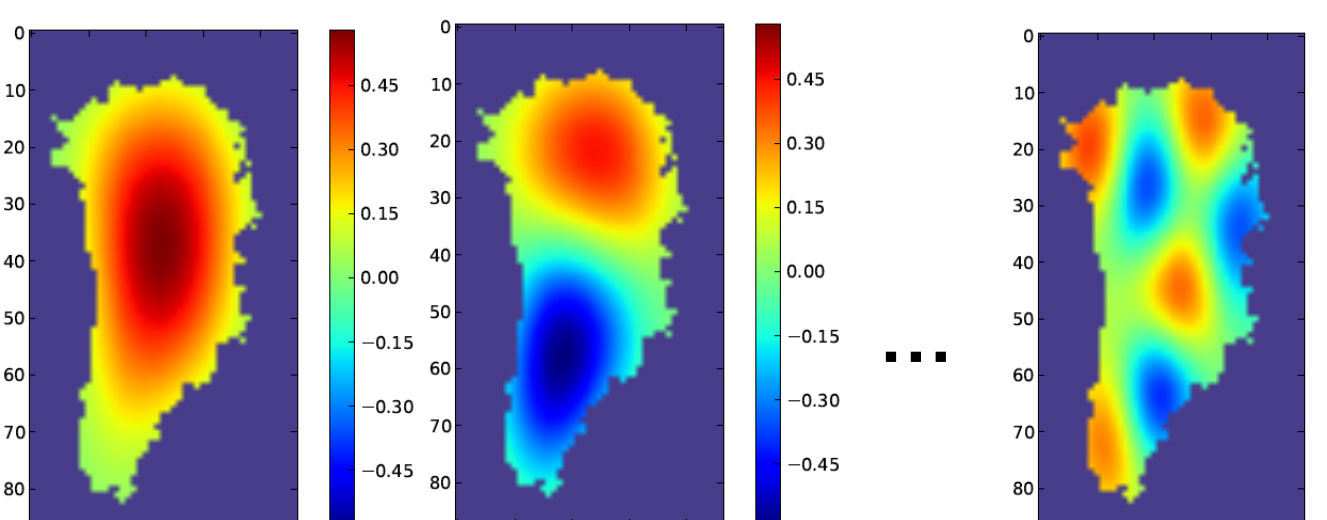
Karhunen-Loève expansion (KLE):

Assume analytic spatial covariance kernel (squared exponential) for random field:

$$C(r_1, r_2) = e^{-(r_1 - r_2)^2 / L^2}$$

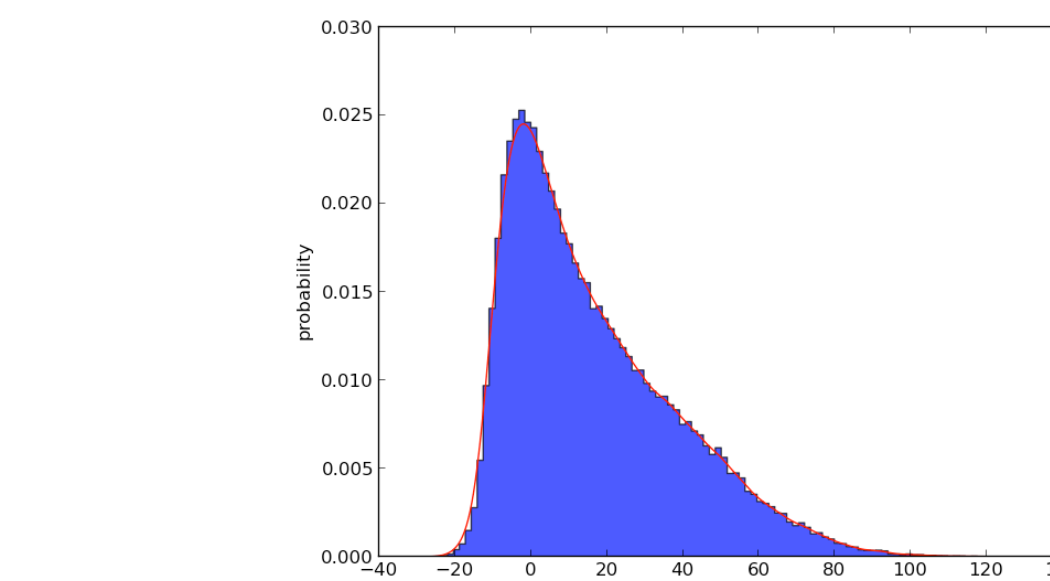
and integrate over domain for modes. Length scale (L) balances feature resolution vs. # KLE modes.

Retain first 10 KLE modes:



Dimension reduced inference and propagation for prediction of SLR:

- Form CS-based PCE emulator using 66 transient solves over prior distributions for 10D KLE parameters



- SLR statistics based on unconstrained priors
- Next steps: emulator-based inference (see Algs poster) → SLR statistics for posteriors constrained by surface velocity data