

The SciDAC FASTMath uncertainty quantification (UQ) team works on development of robust UQ methods within high-quality software, providing SciDAC partnership projects with both production deployments at scale and agile prototyping of tailored capabilities.

Dakota (dakota.sandia.gov)



C++ toolkit that provides a variety of non-intrusive algorithms for design optimization, model calibration, uncertainty quantification, global sensitivity analysis, and parameter studies. 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.

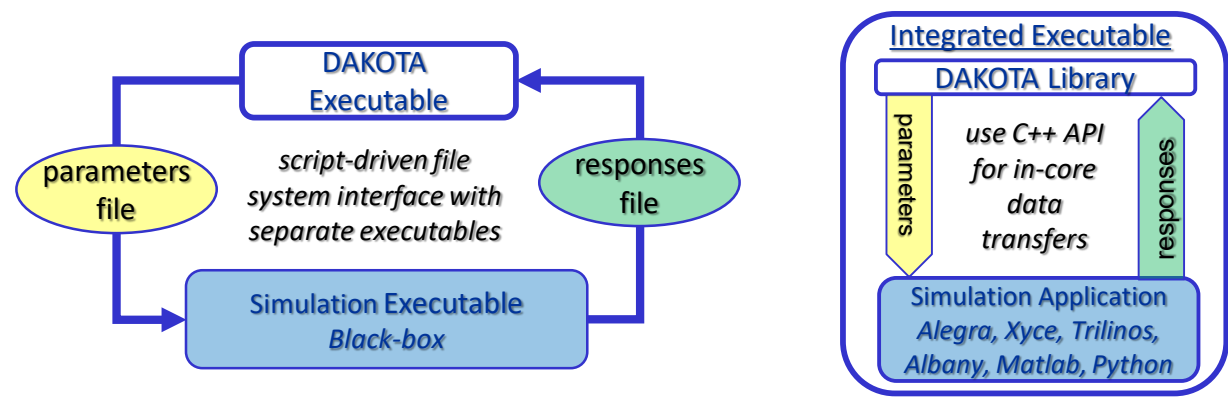
- Capabilities:**
- Core forward UQ components
- Sampling:** Monte Carlo, Latin hypercube; Incremental, Importance, Adaptive
 - Reliability:** Local (FORM, AMV+, TANA/QMEA); Global (EGRA, GPAIS, POF Darts)
 - Stochastic exp.:** PCE (projection, regression); SC (nodal, hierarchical); FTT (see Algs)
 - Epistemic:** Interval estimation (local, global); Dempster-Shafer

- Advanced (multi-component) capabilities
- Bayesian methods:** QUESO, GPMSA, DREAM, MUQ; Emulator-based MCMC
 - Nested studies:** Mixed aleatory-epistemic UQ; Optimization under uncertainty
 - Multilevel-Multifidelity:** sampling (see Algs poster), surrogates
 - Dimension reduction:** Active subspaces; Adapted basis PCE

- Scalable parallelism
- Multilevel parallelism:** MPI + asynchronous local (system call, fork)
 - Asynchronous many task (AMT):** explore ensemble-based UQ workflows with Legion

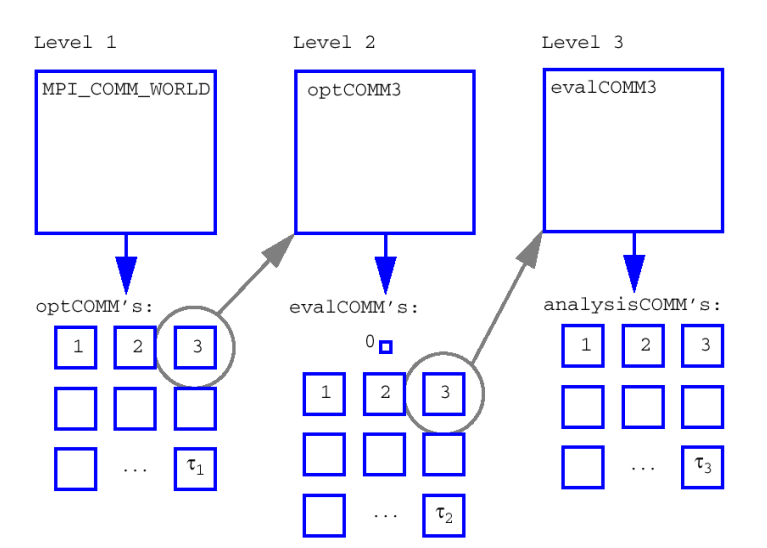
Simulation interfacing

- Black box
- Embedded service

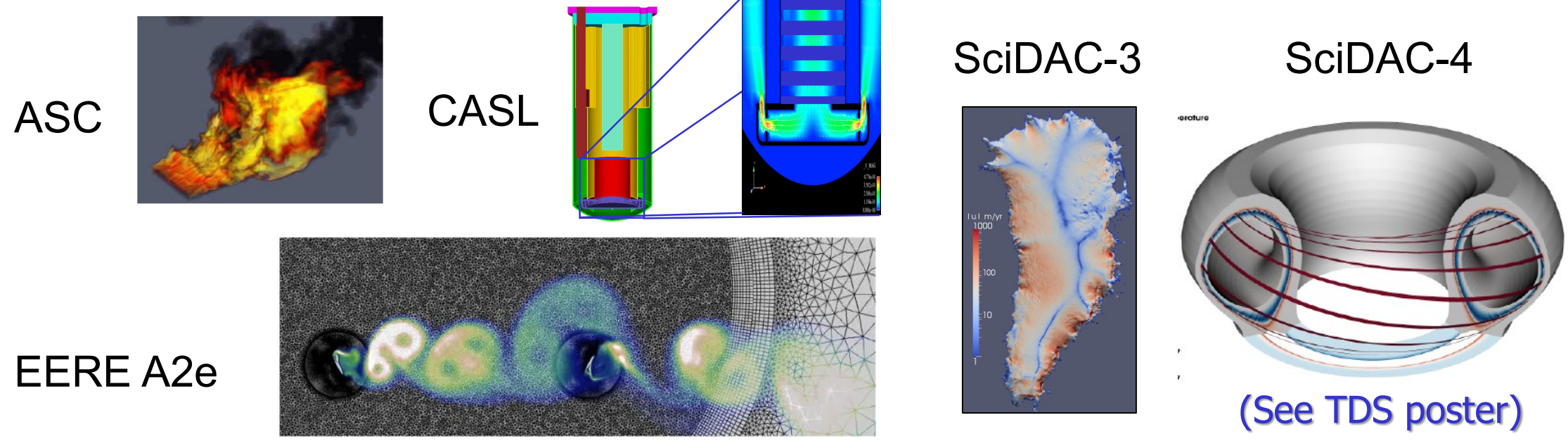


Exploit multiple levels of parallelism

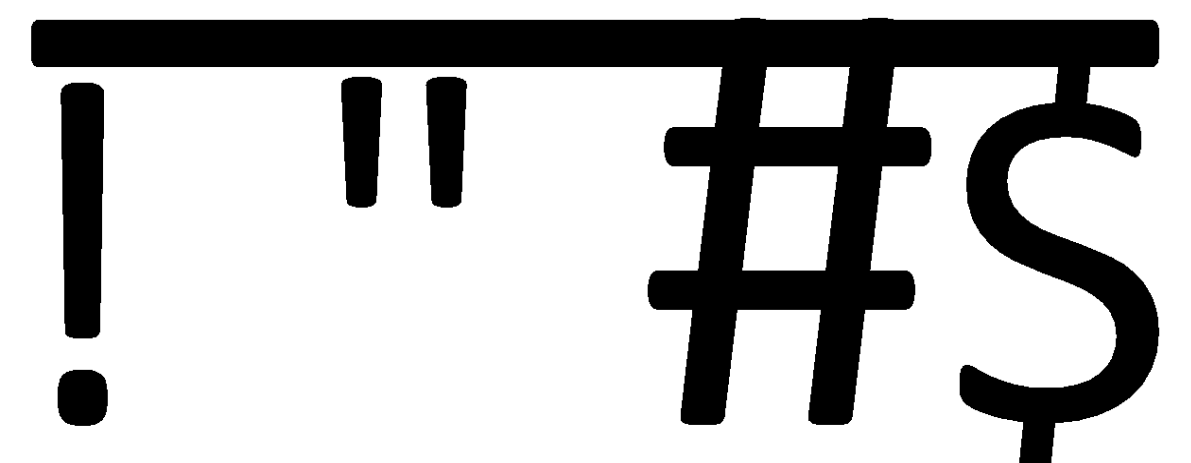
- Recursive partitioning / scheduling with MPI communicators



Defense, Science, and Energy Applications



UQtk (www.sandia.gov/UQToolkit)



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

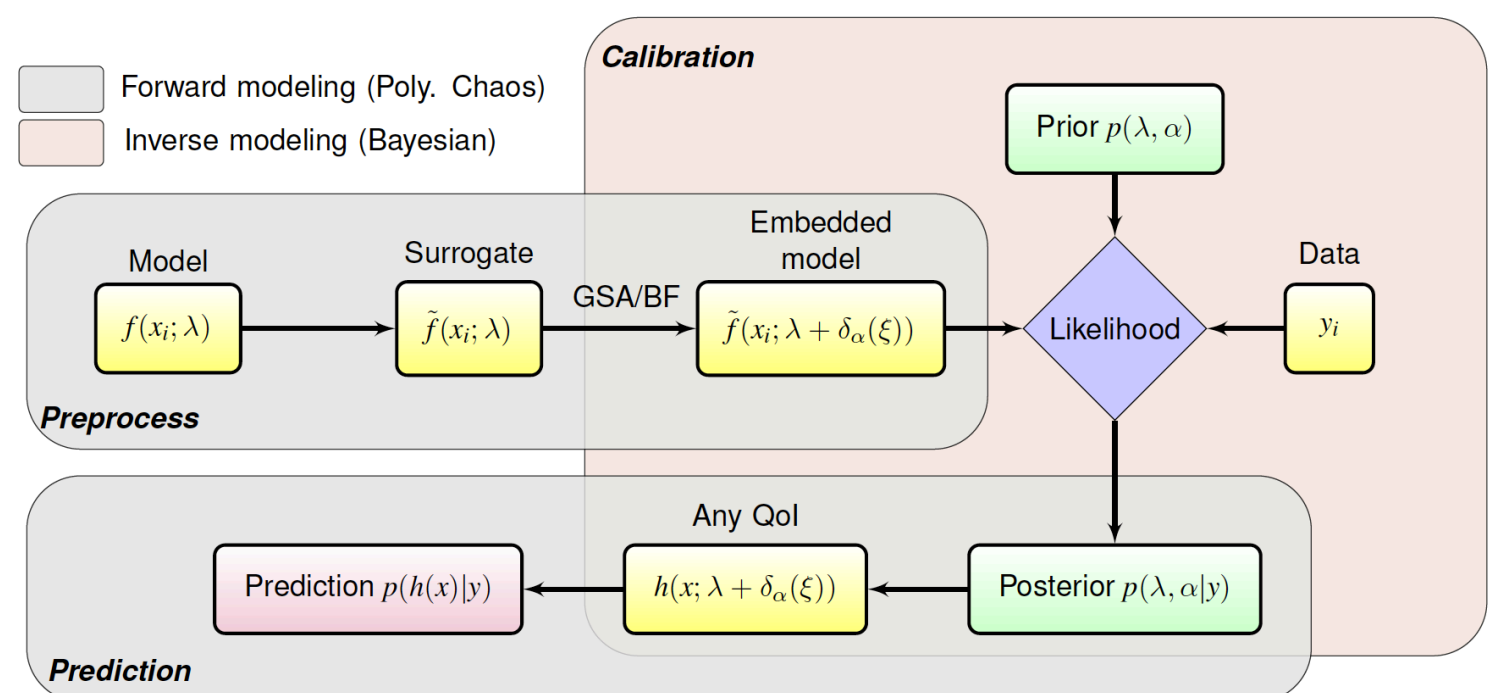
- Complementary to production tools, UQtk targets:
 - Rapid prototyping
 - Algorithmic research
 - Outreach: Tutorials / Educational
- Version 3.0.4 available at <https://github.com/sandialabs/UQtk>
- Version 3.1.0 planned for Fall 2019
- Contact: Bert Debusschere: bjdebus@sandia.gov

Capabilities

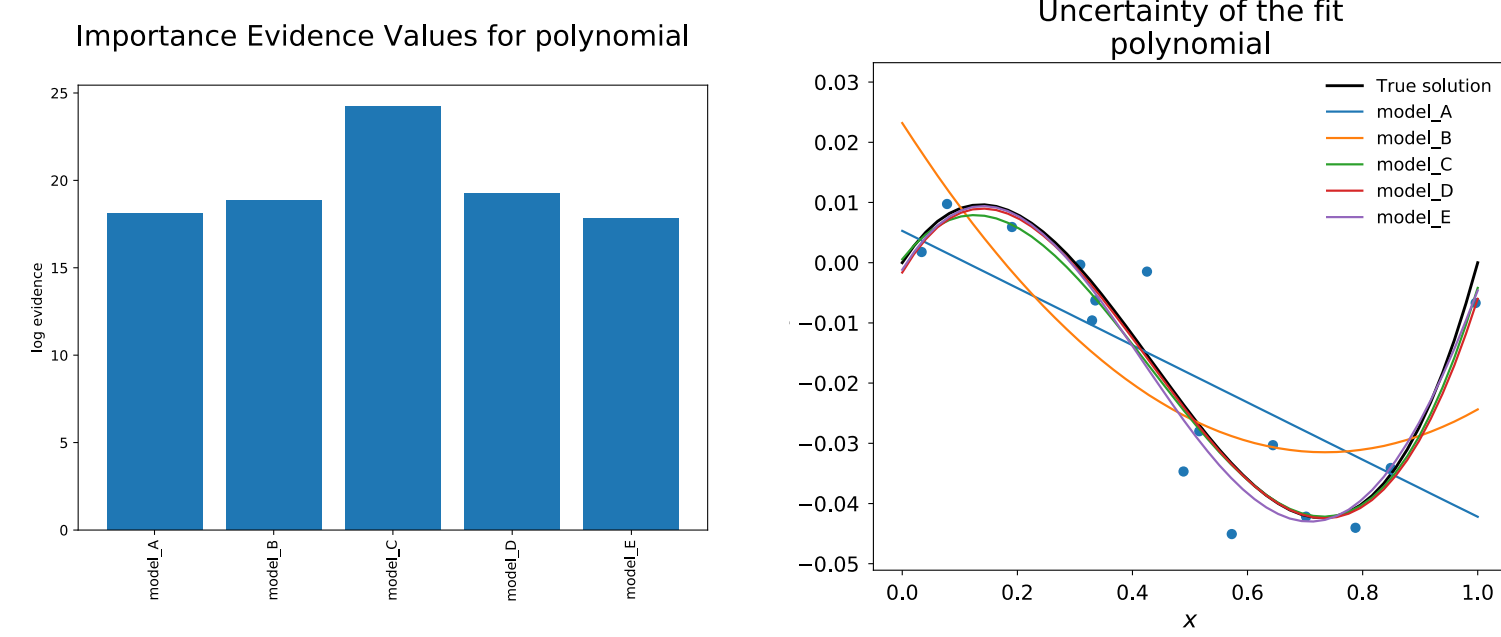
- PC representations of random variables and stochastic processes
- Intrusive and non-intrusive forward propagation
- Bayesian inference with and without model error
- Bayesian Compressive Sensing
- Low Rank Tensors (v3.1.0)
- Data Free Inference (v3.1.0)
- Tools are flexibly combined into comprehensive workflows.

Simulation interfacing

- Direct linking of C++ lib
- Command Line Apps
- Python interface



Bayes Factors for Model Selection



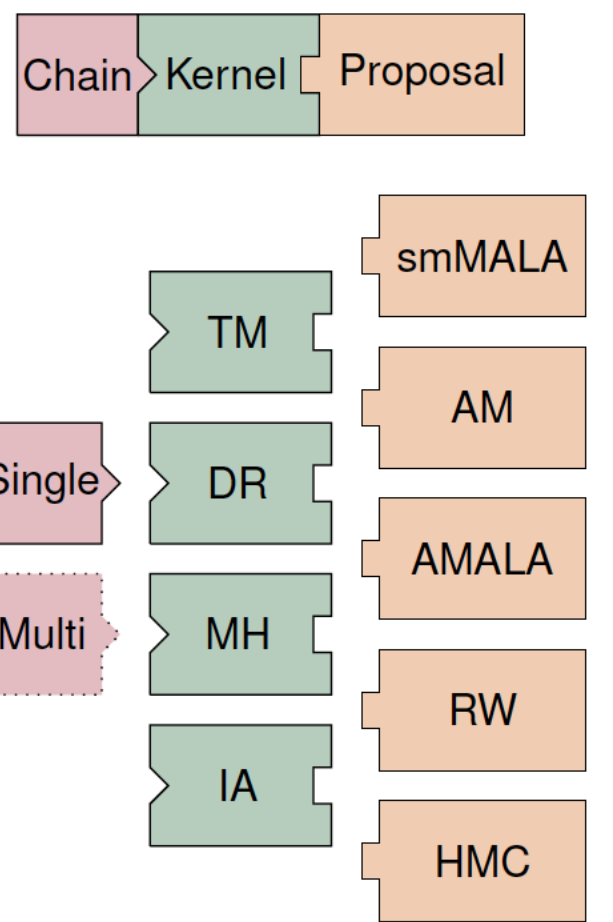
Selected DOE Applications

- BER: OSCM
- FES: PSI-2
- NE: Fission Gas
- BER E3SM
- EERE: HydroGEN

Capability Integration

MUQ

- Capabilities:**
- Bayesian inference for computationally intensive and high-dimensional models, via a suite of advanced Markov chain Monte Carlo algorithms
 - Adaptive surrogate modeling and dimension reduction for scalable inference
 - Transport maps for inference and density estimation/stochastic modeling
 - Graphical framework (DAGs) to describe complex multi-component statistical models, propagating intrusive (e.g., gradient) information when available



- Interfacing with Dakota:**
- MUQ 2.0 integrated as a TPL in Dakota/packages with management of shared dependencies
 - NonDMUQBayesCalibration class defines DAG workflow for inference using MUQ ModPieces

- Interfacing with UQtk:**
- MUQ can readily be integrated in Python UQtk workflows
 - Coupling on C++ library level in progress

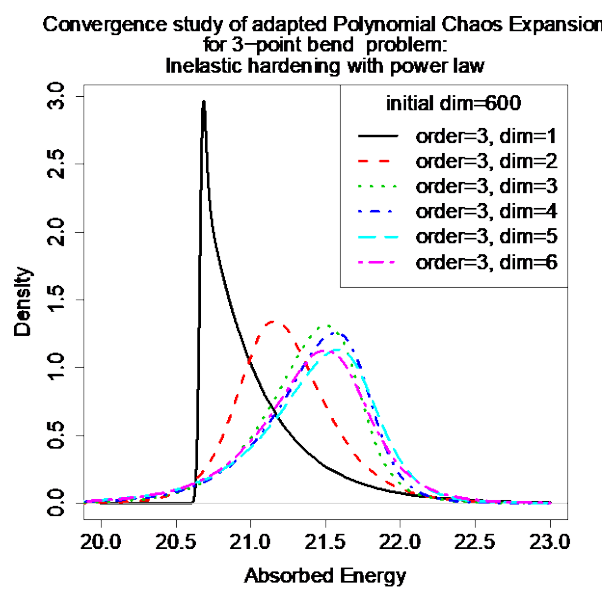
Major components of MCMC: chain, transition kernel, and proposal

Adapted Basis

- Capabilities:**
- Dimension reduction based on polynomial chaos expansion
 - Several algorithms for learning reduction operator, all scale linearly with stochastic dimension
 - Accurate high-order approximation achieved along reduced dimensions
 - Suitable for input parameters with arbitrary joint probability measures
 - Adaptations must be re-learned for each quantity of interest (QoI)

- Interfacing with Dakota:**
- AdaptedBasisModel performs a low-order PCE approximation (for multiple QoI)
 - Rotation defined by first-order PCE
 - Can be consumed by other UQ methods as a recast Model with reduced dimensionality

- Interfacing with UQtk:**
- Iterative scheme discovers converged reduced model
 - Convergence acceleration using posterior error analysis
 - Interface to random field inputs via Karhunen-Loeve expansion



Inelastic impact mechanics with 600 stochastic dimension.