

Data Management and UQ in Extreme Scale XGC Simulation R. Moser¹, M.Greenwald², M. Parashar⁴, S. Klasky³, V. Carey¹, M.Chilenski², S. Ku⁵, J.Choi³ and the rest of EPSI Team U. Texas¹, MIT², ORNL³, Rutgers⁴, PPPL⁵ SDAV A

Uncertainty Quantification

Towards a Bayesian analysis of impurity transport data (MIT) for XGC validation

Objective is to find D, V profiles that best reproduce the observed brightnesses b on each of the diagnostics.

• Nonlinear inverse problem: Key issues are existence, uniqueness and stability of the solution.



ADIOS in EPSI

- Data staging with low-latency, tight coupling execution environments through in-memory data exchanges between different codes.
- Service Oriented Architecture (SOA) for on-demand coupling executions with support of dynamic workflow invocation.







Bayesian Analysis:

- Cubic spline basis functions.
- Adaptive Parallel Tempering(APT) to handle multiple maxima, width of posterior distribution.

X-ray Imaging Crystal and Vacuum UV spectrometer sightlines in Alcator C-Mod



• Agreement on core XICS chords is good in all cases; outer XICS chords shows widest variation • Agreement on VUV spectrometer is reasonable.

- Selection and chunked reads to enable schedule optimization.
- Hierarchical data management with staging for multiple I/O resources and network staging is under development.



Data staging for XGC1-XGCa workflow: In-memory coupling with DataSpaces

XGC

Adios

CADES

ADIOS Vis Schema

- Create an easy-to-use schema for ALL ADIOS codes.
- Facilitates data sharing without adding code complexity
- Visualization schema: Semantics of the data for the purpose of visualization.
- Describing visualization data for various

Data-centric integrated execution environment Our focus is to support EPSI simulation by providing integrated data-centric execution environments for tight code coupling, staged data process, and monitoring system with a support of dynamic workflow system.



- Uncertainty estimate in r/a > 0.6 still too small to be consistent with assumed lack of knowledge there.
- Cases shown are likely overconstrained.
 - **Towards UQ for Extreme Scale XGC Simulations(UT-Austin, PPPL)**

Previous analysis:

Challenges: Global nonlinear system with scale-inseparable and nonlinearly selforganizing multiphysics.

Sensitivity of Plasma Gradients to Applied Heating

- Expense of global gyrokinetic full-f simulation: build surrogate response
- Enrich surrogate(aka response surface) with local sensitivity information
- Perturbational local approach for flexibility, efficiency
- Figure on Right: R/L_t sensitivity to heating power P₀ computed via a global surrogate (curve fit) or local perturbation.
- Local perturbation faster; computable from restart file of "steady-state" high-fidelity XGC1 run(flexibility).
- Currently examining perturb. approach in wedge mode XGC1, XGC1-XGCa coupling.

The UQ "telescoping" approach in EPSI

- 1. UQ in reduced size tokamaks-more details (e.g. gradients) and samples to construct resp. surface 🗸
- 2. Scaling-up the UQ results to larger sizes \checkmark (*telescoping*): limited number of studies
- Calibration/enrichment of the telescoped response surface.
- Validation of predictions against present-day experiments. (including UQ on experimental data). Extension/Numerical version of experimental scaling laws for predicting ITER performance.
- Use of experimental data to inform surrogate model.

- This shows the importance of accounting for the possibility of multiple solutions.
- Next steps: include n_{ρ} , T_{ρ} profile uncertainties.

tools (VTK, Matlab, ParaView).

- XGC Data Management
- Hierarchical data Manage hierarchical data and resources management
- Integrated staging services
- Support for ADIOS Vis Schema.



EAVL: Extreme Scale Analysis and Visualization Library

XGCa

XGC1

EPSI coupling workflow

Goal: To enable tightly coupled XGC1 and XGCa workflow using memory to memory coupling.

DataSpaces As a Service

- Provides data staging as a service to applications persistent data and staging services across application instances
- Allows XGC processes to dynamically connect/disconnect to/from the staging service
- New approach targets more complex and dynamic workflows with tighter coupling; provides more flexibility
- DataSpaces as a Service improves resource efficiency and increase I/O performance

Illustration of using data staging service (DataSpaces-as-a-Service) to build the XGC1-XGCa workflow: connecting the different component applications, enabling data exchange and sharing.

DataSpaces As a Service

XGC1

Timeline

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XGCa





3. Telescope further to ITER-scale, compare prediction against ITER-scale XGC1 simulations

Telescoping and calibration considerations:

- Simulation and data-driven separation of scale-independent and -dependent quantities.
- Physics guidance important-response surface should inform, not dictate, higherfidelity studies.
- Negative telescoping results also useful-identify key regimes in parameter space for high-fidelity simulation(e.g. bifurcations in parameter space), compatible with Expected Information Gain(EIG) base approaches. **Initial "Telescoping" Studies Details**
- Builds on mod. CYCLONE heating study-provide Q, dQ/dP at various heating power
- Telescoping parameter is $1/\rho^*$ (or a/ρ_i). Scaling is achieved by increasing magnetic field
- Compute Q,dQ/dP at range of ρ^* values, as well as multiple values of P₀.
- Compute krieging surface in the scaling parameter and the uncertain model • parameter(s).
- Predict Q,dQ/dP at a higher value of a/p_i . How valid is "reduced-size" data? •
- As number of model parameters increase, leverage QUEST software: QUESO, GPMSA



"Steady-state" gradient sensitivities with

respect to heating power computed on a

Edison and Hopper.

Physics

UQ telescoping

high temperature version of the CYCLONE

base case. Simulations performed on NERSC

Future "Telescoping" Work

- Approach validated in for $.5 < \rho^* < 1.5$ CYCLONE case values.
- Adding additional QoI(thermal diffusivity)
- Add additional parameters present in heating model (torque, source profiles) • Incorporate uncertainty in magnetic data into prediction response surface.
- Derive approx. uncertainty estimates for future use in response surfaces.
- Move response surface simulations and scaling closer to available exp. data.

Data-centric Mapping in DataSpaces

App2

computer node 2

with 12 cores

Graph of data transfers between coupled codes

Data Staging

coupled region distributed coupled region distributed over 20 processes over 4 processes

App1

5410

17 16 13 12

computer node

1 with 12 cores



Locality-aware placement of coupled region processes on Cray XT5 12-cores computer nodes App1 process n this example: App2 process (1) Inter-node coupling data transfers (over networks) is minimized

Illustration of the data-centric mapping of the application process for

concurrently coupled workflow

(2) Intra-node coupling data transfers can be performed using more efficient shared memory

File-based File-based Server-based Server-based App1 process App2 process 1.24 0.12 Number of simulation processes (cores) Number of simulation processes (cores)

Preliminary Results for In-node Staging

- Performance comparison with file-based staging and in-memory staging on dedicated servers
- Particle data read time reduced by ~98% compared to file-based and by ~90% compared to staging on dedicated servers
- Turbulence data read time reduced by ~99% compared to file-based and by ~96% compared to staging on dedicated servers