

Mathematical Challenges in Multiscale Approaches and Related issues in Nuclear Reactor Simulation

Mihai Anitescu, Argonne

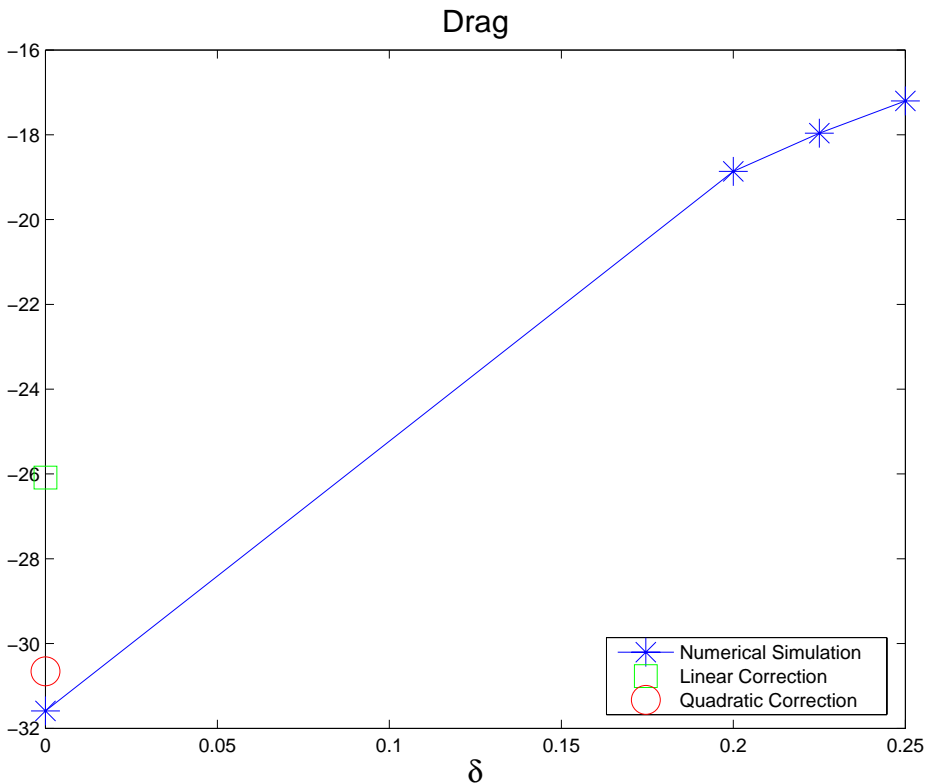
August 16, 2006



Structure of talk

- Multiscale example problems examples and challenges (hopefully) relevant for GNEP, based on the speaker's (small) experience.
- General challenges that can be abstracted from these examples?

Parametric Sensitivity of Large Eddy Simulation (LES)



2d drag on fixed cylindrical body

- A subgrid model for fluid flow seems unavoidable, so its effect on the “reality” should be quantified.
- Parametric sensitivity (Smagorinski, left panel) of LES leads to improved estimates of flow functionals, compared to LES itself (with William Layton, Pitt)
- **Our finding:** For accurate estimation, the sensitivity needs to be estimated on a finer mesh than LES, but **coarser than the one needed for direct numerical simulation.**
- **Open questions:** Other LES, comprehensive error analysis, a posteriori error analysis

Multiscale Approaches for Problems in Material Science.

- ... inspired by the quasicontinuum approach (Tadmor et al.).
- High resolution model

$$(O) \quad \begin{array}{ll} \min & f(x_1, x_2) \\ \text{sbj. to} & g(x_1, x_2) = 0. \end{array}$$

- Representative (coarse-scale) DOF, x_1 , $\dim(x_1) \ll \dim(x_2)$.
- Key observation: at the solution of the problem we have $x_2 \approx Tx_1$, where T is an interpolation operator.
- Replace (O) with (RE) , of much smaller dimension, by writing optimality conditions and using the interpolation rule.

$$(RE) \quad \begin{array}{ll} \nabla_{x_1} f(x_1, Tx_1) + \nabla_{x_1} g(x_1, Tx_1) \lambda & = 0, \\ g(x_1, Tx_1) & = 0. \end{array}$$

Fundamental question: is the reduced problem well posed ?/

- **Yes** (Anitescu, et al., 2006) ... under certain assumptions. There must be a certain local compatibility between the interpolation operator and the energy (or Lagrangian) functional.

$$x = (x_1, x_2), \quad L(x, \lambda) = f(x) + \lambda g(x)$$

$$\left\| \nabla_{x_2 x_2}^2 L(x^*, \lambda^*) T + \nabla_{x_2 x_1}^2 L(x^*, \lambda^*) \right\| \ll 1$$

- Related question: Is it better to Optimize and Interpolate (force based) or Interpolated and Optimize (energy based).

$$(RO) \quad \min \quad f(x_1, Tx_1) \\ \text{sbj. to} \quad g(x_1, Tx_1) = 0$$

- Answer: the latter, but its data are harder to compute.
- **Open questions: Improved reduction of Interpolate and Optimize, the case of inequality constraints, a posteriori error estimation, efficient solvers for the subproblems.**

A multiscale approach for orbital free density functional theory

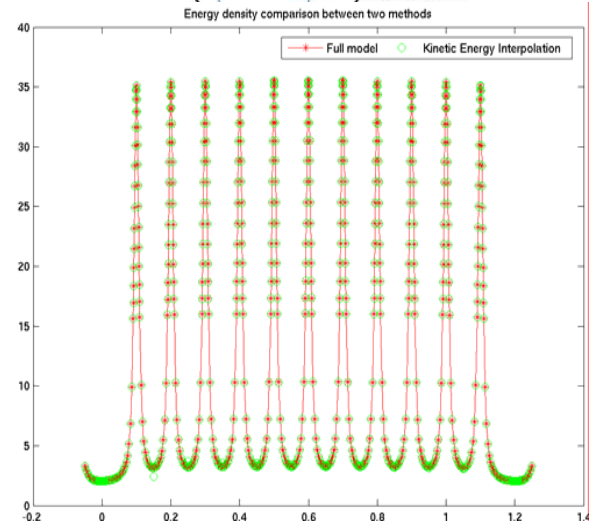
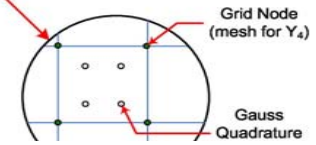
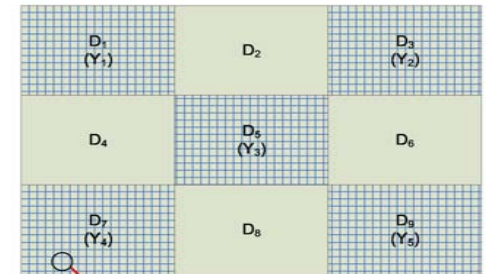
- Representative variables: The density in the representative domains

$$Y_\alpha, \quad \alpha = 1, 2, \dots, P$$

- The interpolation operator is constructed with respect to a reference crystalline mesh (2005, in print)

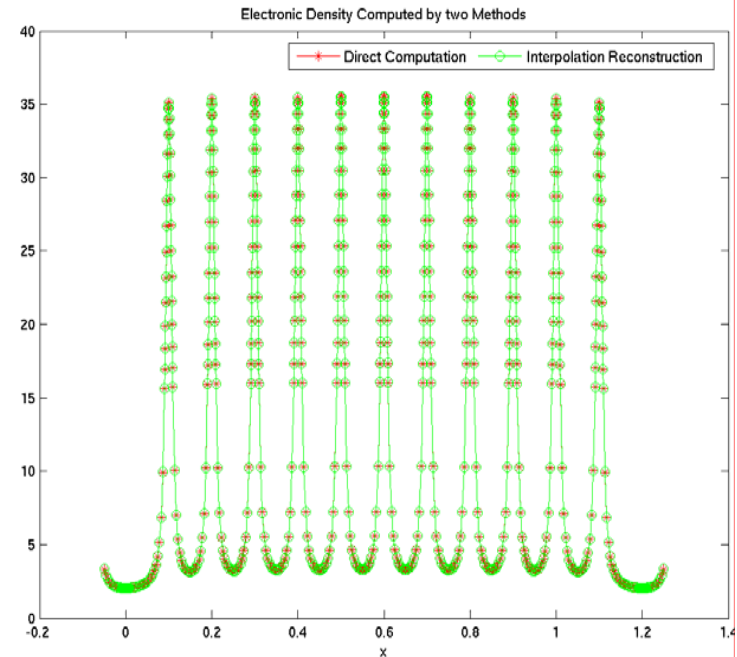
$$\rho_i(\Phi(\mathbf{r}^0, t)) = \sum_{\alpha=1}^p \mathcal{I}_\alpha(i) \rho_\alpha(\Phi(\mathbf{r}^0 + \mathbf{T}_{i\alpha}, t)).$$

- In material problems appearing in radiation damage, accurate potentials are not available so DFT is an essential.
- We have proved that the approach is well posed.



How well does it do?

- Parallel solver, based on PETSC/TAO, created in the last 3 months.
- Example for 11 hydrogen atoms, 1D and 3D
- Open questions:
 - How to extend to the case the energy functional is not explicit? (Kohn-Sham)
 - Efficient ways to compute, store and use the reduced Hessian matrix?
 - What are useful optimization algorithms that exhibit minimal communication (the domain decomposition angle? What are appropriate boundary conditions?)



RelativeError

0.000900 4.48 8.97 13.5 17.9



General open questions applicable to multiscale model reduction.

- Parametric sensitivity equations and a posteriori error estimation for homogeneous multiscale method reduction.
- What are consistent model reduction of and efficient algorithms for multiscale models with constraints (e.g. total charge and nonnegativity constraints), when only representative subdomain solves are considered?
- What is a consistent mathematical framework for multiscale model reduction, which is useful for uncertainty quantification? For example deterministic high resolution deterministic problem reduced to stochastic differential equation (e.g. molecular dynamics -> Langevin, with high resolution solves to compute the reduced parameters).
- Can this framework be used to define a domain-specific language for nuclear reactors that would accelerate component integration and HPC performance (freeFEM and Sundance for FEM). That is essential for efficient development of future workforce.
- For back-end sensitivity calculations, how does one do accelerated sampling? If answer is randomized quasi Monte Carlo, can we prove convergence in distribution that is essential in statistical estimation?