Implementing hybrid PDE solvers

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Outline

Objectives

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Basic Implementations

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Synopsis and prospects

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Overall Objective

Effectively combine

- conventional deterministic PDE solving methods and
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for solving linear Elliptic Partial Differential Equations.

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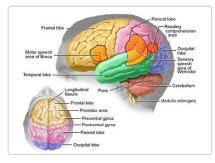
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Long term goal

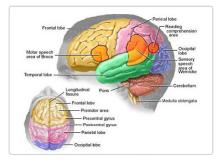
Provide practical (production?) computational tools that add value to existing high performance PDE solvers.

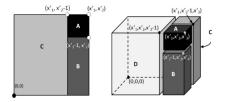
Stochastic pre-processing A Monte Carlo-based walk on spheres approach is utilized to decouple the original PDE problem into a set of independent PDE sub-problems. Deterministic solving Each of the resulting sub-problems is solved independently by an appropriately selected PDE solver.

Long and short term: PDE problems

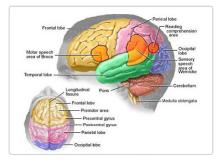


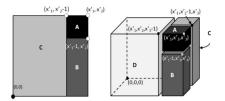
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BEWARE:

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Specific Algorithm (1/2)

PDE Problem

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$$Bu(x) = g(x) \quad x \in \partial \mathcal{D},$$

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Subdomains and Interfaces

$$\mathcal{D} = \cup_{\mu=1}^{\mathcal{N}_{\mathcal{D}}} \mathcal{D}_{\mu}$$

 $\mathcal{I}_{\mu,\nu} = \partial \mathcal{D}_{\mu} \cap (\partial \mathcal{D}_{\nu} \cup \mathcal{D}_{\nu}) \subset \mathbb{R}^{d-1}, \quad \mu \neq \nu, \quad \mu, \nu = 1, \dots, \mathcal{N}_{\mathcal{D}}.$

Specific Algorithm (2/2)

Data: i_1, i_2, \ldots, i_N : subdomains we wish to compute the solution. **Result**: \tilde{u}_{μ} , $\mu = i_1, \ldots, i_N$: computed solutions in \mathcal{D}_{μ}

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// PHASE II: Compute solution in subdomains for j = 1, 2, ..., N do Solve the PDE problem:; $L_{i_j} u_{i_j}(x) = f_{i_j}(x) \quad x \in \mathcal{D}_{i_j}$; $B_{i_j} u_{i_j}(x) = g_{i_j}(x) \quad x \in \partial \mathcal{D}_{i_j} \cap \partial \mathcal{D}$; $L_{i_j}^* u_{i_j}(x) = h_{i_j}(x) \quad x \in \mathcal{D}_{i_j}$

end

State-of-the-art

- M. Muller, Some Continuous Monte Carlo Methods for the Dirichlet Problem, Annals Mathem Statistics, 1956.
- J. DeLaurentis and L. Romero, A Monte Carlo method for Poisson's equation, J. Comput. Phys., 1990.
- M. Mascagni, A. Karaivanova, and Y. Li, A quasi-Monte Carlo method for elliptic PDEs. Monte Carlo Methods and Applications, 2001.
- R. Papancheva, I. Dimov, and T. Gurov. A new class of grid-free Monte Carlo algorithms for elliptic BVP, Numerical Methods and Applications, 2003.
- J. Acebron, M. Busico, P. Lanucara, and R. Spigler, Domain decomposition solution of elliptic boundary-value problems via Monte Carlo and quasi-Monte Carlo methods. SIAM Journal of Sci. Comp., 2006.
- PDD-HPC Research project (Acebron at al.)

Nevertheless

To the best of our knowledge

- ► No systematic experimentation as been performed.
- No state-of-the-art PDE solving modulus have been considered.
- No practical issues have been raised.
- No software components are readily available.
- ► No new computing paradigms have been explored.

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Isolated and sporadic efforts.

Basic algorithms and software components

- Our QMC implementations based on known random walks on spheres algorithms. GRID FREE!
- Use of standard C++ library
- Common 2D and 3D user interface
- SINTEF's Multilevel B-splines library for interpolation
- State of the art PDE solvers from http://www.dealii.org/ (2007 J. H. Wilkinson Prize for Numerical Software)
- State of the art graphics integration (TecPlot)

Selected experiments in 2D

$$\begin{aligned} &\frac{\partial^2 u}{\partial x^2} + \frac{\partial^2 u}{\partial y^2} = f(x, y), \ \forall (x, y) \in \Omega \equiv [-1, 1] \times [-1, 1], \\ &u(\pm 1, y) = \quad \cosh(\pm 2) \cos(2\pi y) \\ &u(x, \pm 1) = \quad \sin(\pi x) \sinh(\pm 1) + \cosh(2x), \end{aligned}$$

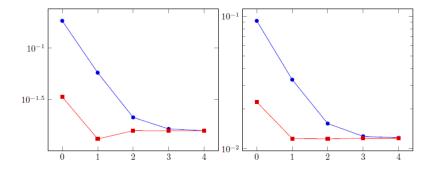
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- ▶ 8 subdomains with interfaces at x₁ = 0, y₁ = -0.5, y₂ = 0 and y₃ = 0.75.
- Seek the solution only in $\Omega_{1,0}$, $\Omega_{0,1}$ and $\Omega_{2,1}$

Error reductions for various configurations



GPU experimentation

Computing devices:

- CPU: Intel(R) Core(TM) i7 CPU 870 (2.93GHz)
- ► GPU: GeForce GTX 480 (1401MHz)
- Speedups CPU/GPU+CPU
 - Maximum speedup of QMC: 13x
 - Maximum overall speedup: 150x

Trivial implementation!

Web Services experimentation

Computing servers:

- Local PCs
- Tier 3 local computational servers
- Amazon virtual machines

Low communication/computation ratio

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New line of reasoning that provides new intuition about the dynamics of PDE MC simulations.

https://github.com/mvavalis/Hybrid-numerical-PDE-solvers