

Multi-level Optimization Algorithms for PDE-constrained Problems Arising in the Design of Nanoporous Materials

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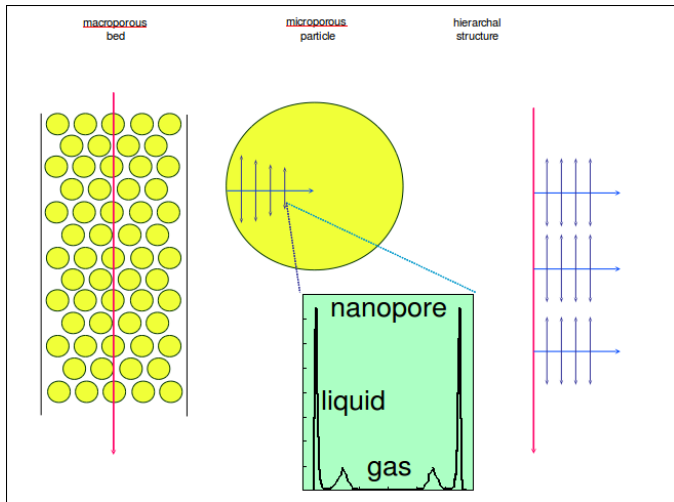
Outline

- 1 The Nanoporous Materials Problem
- 2 The Multi-Level (Hierarchical) Optimization Problem
- 3 The Multi-Grid Optimization Algorithm
- 4 Progress and Results
- 5 Future Work

- David Gay (AMPL Corp.)
- Stewart Griffiths (Sandia National Labs)
- R. Michael Lewis (College of William and Mary)
- Kevin Long (Texas Tech University)
- Stephen Nash (George Mason University)
- Robert Nilson (Sandia National Labs)

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Nanoporous Materials



Nanoporous Materials Promise

- Within nanopores, can store gas at high density with low pressure
 - Gas liquefies at ambient temperatures when very close to surface
 - In nanopores all molecules are close to surface
- By similar mechanism, ions pack at higher density in nanopores
 - Nanoporous materials can be used for capacitors or batteries

Nanoporous Materials Design: A Multi-Scale Physics Problem

- Consider gas storage
- Trade-off: get gas in and out quickly vs. maximize storage
 - Improve the flow by making channels in the material, but this lowers storage
- Physics changes as the scale is reduced
 - Make channels of smaller widths at finer scales
- Must get physics correct on all levels

Nanoporous Materials Design: A Multi-Scale Physics Problem

- Hierarchical optimization model with multi-physics
- Can't solve problem on fine level directly (too big)
- Can we exploit the structure to create efficient algorithms?
- Issues:
 - How many levels to use
 - How to model the channels
 - How to aggregate at various levels
 - How to exploit parallelism at small scales
 - What algorithmic framework to use
 - How to construct initial guess

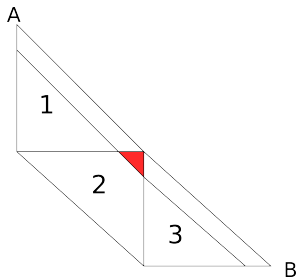
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The Multi-Level (Hierarchical) Optimization Problem

- The flow is described by a nonlinear, time-dependent, PDE on the coarse levels, but the parameters change as scale decreases
- It is a molecular dynamics problem on the fine level
- The physicists have shown that good steady-state approximations can be constructed

Modeling the Channels



- Allow flow along the edges of the finite element mesh
- Let w_i be the width on edge with unit tangential vector e_i
- Derive an effective mobility tensor $B = \gamma w_i^2 e_i e_i^T$ for constant γ
- Total flow is flow in channel combined with flow in the bulk

First Problem: Steady State

- Let $\chi(x)$ be 0 in the bulk and 1 in the channels
- The net mass transfer is

$$j = -\chi(x)\rho_1 B \nabla p - (1 - \chi(x)\rho_2 \nabla p$$

- The transport equation is

$$c + \nabla \cdot j = 0$$

where c is a constant

Finite Element Equation

- Create the weak form of the PDE with nondimensional constants Q , D , and V and test function $\bar{u}$

$$\begin{aligned} & \int_{\Omega} [Q\bar{u} + D\nabla\bar{u} \cdot \nabla p] \xi \, dV \\ & + \sum_{i=1}^N \int_e a_i \left[V\rho \frac{\partial\bar{u}}{\partial l} \frac{\partial p}{\partial l} \right] dl \\ & + \int_{\partial\Omega} \bar{u}\mathbf{j} \cdot \mathbf{n} \, dS = 0 \end{aligned}$$

- A difficulty arises because typical values for the constants are:

$$Q = 1$$

$$D = 2$$

$$V = 10^{12}$$

The Optimization Problem

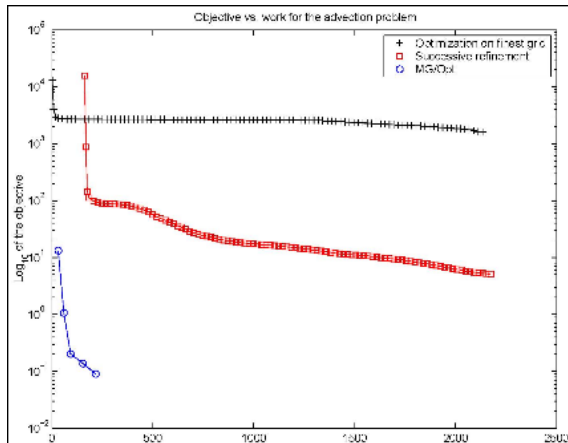
- The objective function depends on the application
- An example is to minimize mean pressure

$$f(w) = \int_{\Omega} p(w)$$

- Subject to the PDE

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Why Multi-Grid Optimization Methods



Three different ways to apply the *same* optimization algorithm:

- straightforward
- successive refinement
- multilevel

The Multi-Grid Optimization Algorithm

- Notation (Consider only two levels and unconstrained optimization)
 - Let x_h be the variables on fine level h
 - Let x_H be the variables on coarse level H
 - Let I_h^H be the “downdate” operator and I_H^h be the “update” operator to transform between levels
 - Let x_h^0 be the initial guess on the fine level
 - Let f_h and f_H denote the function to be minimized at each level

The Multi-Grid Optimization Algorithm: MgOpt

■ Pre-smooth

- If on the coarsest grid, solve completely and return
- If not, perform $k_1 > 0$ iterations to obtain x_h

■ Recursion

- Compute $\bar{x}_H = I_h^H \bar{x}_h$ and
 $\bar{v}_H = \nabla f_H(\bar{x}_H) - I_h^H \nabla f_h(\bar{x}_h)$
- Call MgOpt on the "surrogate" model

$$f_s \equiv f_H(x_H) - \bar{v}_H^T x_H$$

to obtain x_H^+

- Compute the search directions $e_H = x_H^+ - \bar{x}_H$ and $e_h = I_h^h e_H$
- Do linesearch to obtain α such that $f_h(\bar{x}_h + \alpha e_h) \leq f(\bar{x}_h)$

■ Post-smooth

- Apply k_2 iterations to obtain x_h^{j+1}

Algorithm (Software) Design

- We want a general form of the MgOpt algorithm
 - Algorithm should be independent of the specifics, i.e., it shouldn't care about constraints, optimization strategies on each level, etc.
- Thus we created a “Level” object that knows
 - Objective function
 - Constraints (if any)
 - Merit function
 - Update and downdate operators
 - Computation of the search direction
 - Optimizer, including linesearch and convergence
- We construct a set of these to describe all the levels
- We allow various forms of the algorithm:
 - Can start at fine level or coarse level, use “full multi-grid” etc.

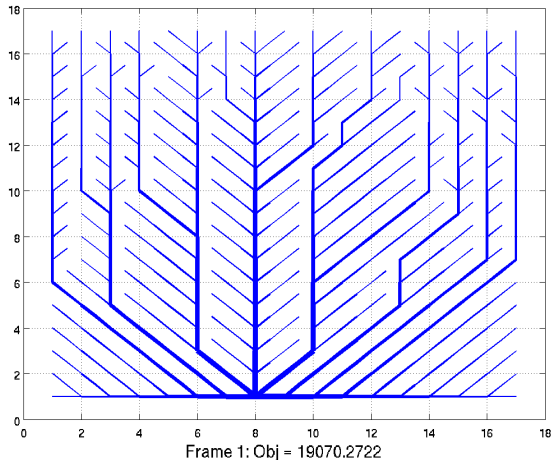
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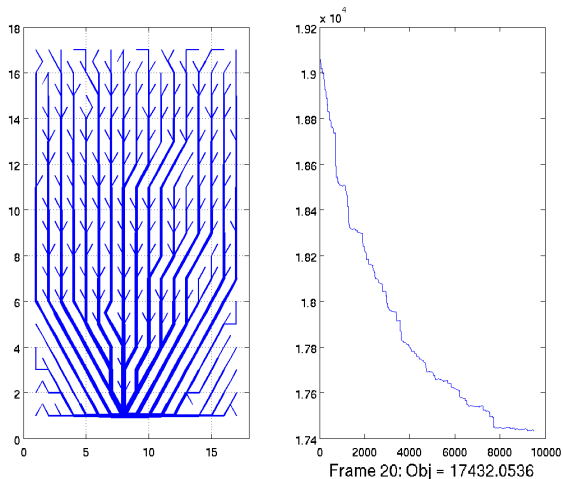
Progress and Results: Initial Guess

- We used a network approximation based on the finite element grid
 - Source of 1 at each vertex
 - Each edge is an allowable channel
 - Used a greedy algorithm to get initial set of edges with non-zero flow
 - Used a Metropolis-Hastings algorithm to improve flow

Progress and Results: Initial Guess with Greedy Algorithm



Progress and Results: Initial Guess Improved by MH



Progress and Results: Problem Analysis

- Created accurate steady-state approximations
- Created a simple problem on simple grid to test procedures
 - Can check with both finite difference and finite element methods
 - In some cases we can get analytic solutions
- Developed specific update procedures: if 2 edges become one on the coarser level, compute the harmonic average to preserve flow

Progress and Results: Algorithm and Software

- We have derived a basic convergence analysis of MgOpt with constraints (Nash)
- We have a basic version of the software running
 - Uses Sundance to handle the PDEs
 - Have run a 1-d Laplacian on 4 levels to debug
 - Have run the basic nanoporous problem on one level to debug the PDE implementation

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- We developed a strategy for grid refinement; we want to do it automatically, not by hand (Kevin Long is working on this)
- Continue creating test problems with added features
- Investigate strategies for handling constraints, especially inequality constraints
- Model “real” applications

¡Gracias!