

# Combining Multi-Grid and Domain Decomposition as Preconditioners for a Class of Multi-Level PDE-Constrained Optimization Problems

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- 1 The Domain Decomposition Optimization (DD/Opt) Algorithm
- 2 Status and Future Work

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- Kevin Long (Texas Tech University)
- Stephen Nash (George Mason University)
- Julien Cortial (Sandia National Laboratories)

- 1 The Domain Decomposition Optimization (DD/Opt) Algorithm
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# Domain Decomposition to Extend a Hierarchical (Multi-Grid) Optimization Algorithm

- In our problems, we have a natural hierarchy of levels with the physics described by a PDE at each level
- The optimize problem is formulated on the finest level
- We use the coarser levels in a multi-grid optimization (MG/Opt) algorithm (speeds convergence)
- To use MG/Opt, we have to iterate on finest level, but it will be too large, i.e., the problem is at the nano scale

We have developed a preliminary DD/Opt strategy that blends with the MG/Opt algorithm to extend the range of applicability

Lots left to do

Picture of domain and subdomains, including overlap

# Problems and Domains

- Let  $\Omega$  be the whole domain and  $\Omega_i$  the  $i^{th}$  subdomain
- Assume that there are  $S$  subdomains
- Let  $f$  be objective function on  $\Omega$  and  $f_i$  the objective on  $\Omega_i$
- Let  $x$  be all of the variables on  $\Omega$  and  $x_i$  the variables on  $\Omega_i$
- Let  $x^0$  be the initial guess

# The Domain Decomposition Optimization (DD/Opt) Algorithm for Unconstrained Problem

Set  $n = 0$

- 1 Extract  $x_i^n$  from  $x^n$  for each  $i$
- 2 Perform  $k_1$  iterations on each subproblem  $i$  to obtain  $x_i^{k_1}$ ; compute the local search direction  $d_i = x_i^{k_1} - x_i^n$
- 3 Assemble the search direction into global search direction  $d = (d_1, \dots, d_S)^T$  (handling the overlap is an issue)
- 4 Perform  $k_2$  iterations on

$$\min_s \frac{1}{2} s^T B s + s^T d,$$

where  $B$  is an approximation to  $\nabla^2 f(x^n)$

- 5 Perform a linesearch on  $f(x^n + \alpha s)$
- 6 Check convergence;  $n := n + 1$ ; repeat as necessary



# Comments on the Algorithm

- If  $k_1 = 0$  then the direction  $d_i$  is taken to be  $-\nabla f(x_i^n)$ , i.e., steepest descent
- If  $k_1 \neq 0$  then  $d_i$  can be considered to be a *preconditioned gradient*
- If  $k_2 = 0$  then  $s = d$
- If  $k_2 \neq 0$  then we can solve with a preconditioned CG method using the subdomain solves as the preconditioner

# Issues Being Investigated

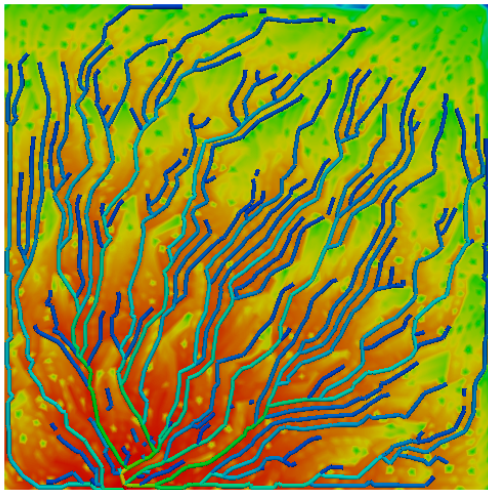
- How to set up and solve the subdomain problems  
e.g., set subdomain BCs, handle the overlap
- How many iterations (i.e.,  $k_1$  and  $k_2$ )
- How to iterate on the full problem
  - Need to evaluate the function and gradient, i.e., solve PDE for fixed value of design variables
  - Use DD along with MG/Opt to facilitate this evaluation
- How to deal with the constraints
  - Step 4 becomes a quadratic programming problem
  - Bounds have to be shifted
  - Other constraints have to be distributed

# The Motivating Problem

(See next talk by Dave Gay for the details)

- We have a PDE-Constrained optimization problem
- Motivated by gas storage in a nanoporous material
- Want to put channels into material to facilitate charge/discharge of the gas

# A Channel Network in a Nanoporous Material



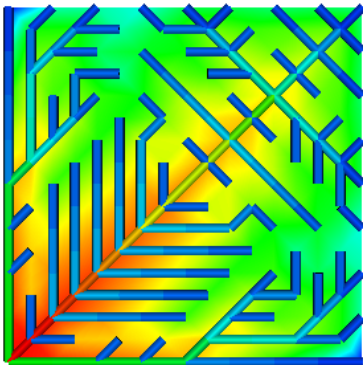
# The Motivating Problem

- Trade-off between speed of charge and volume of material devoted to channels
- Our model allows flow along edges of finite element grid
- We want wide main channels and fine subchannels
- We create a hierarchy of subgrids to effect this

# The Motivating Problem

- There will be many scales from macro to nano
- The physics will change as we refine the scale
- Can pose problem on finest scale, but it will be very large
- Question: Can we use the structure to create more efficient algorithm?

# Example Showing Tree Structure



# The Tree Constraint

- We use a network approximation to the physics to obtain a starting guess
- Can show that the optimal network will be a tree structure
- Thus, we want to maintain a tree constraint  
(This, of course, is an integer constraint)
- Our MG/Opt procedure starts with a tree and then “locks” into a tree after a small number of iterations



# The Tree Constraint

- Hard to maintain tree structure in DD/Opt
  - Can't just do it locally because of boundary problems, i.e., disconnects and occur across boundaries on the subdomains
- A subdomain may see no need for an edge to be maintained
- We developed a global “treeifying” procedure that settles on a tree by fast, local method, i.e., it can be parallelized

# Some Results

(Will put in some results soon)

- Problem
- size
- solve by lmbfgs
- solve by DD
- solve in parallel

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- Have paper describing  $\chi Opt$  accepted in SIOPT
- We have derived a basic convergence analysis of  $\chi Opt$  with constraints (Nash)
- We have a reasonable implementation of the software running
  - Uses Sundance to handle the PDEs
  - Have run a variety of tests
  - Have designed and implemented an automatic mesh refinement algorithm
  - Have coded the special update and downdate procedures making use of automatic refinement
  - Have added strategies to handle the “integer” nature of the problem, including the initial guess
  - Have developed DD/Opt that can be incorporated in this framework

- Investigate strategies for handling constraints, especially inequality constraints, e.g., arising from manufacturability requirements
- Incorporate DD/Opt into  $\chi Opt$

# Thanks!

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