

AD AND COMBINATORICS: AN ACCIDENTAL PARTNERSHIP?

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Joint work with
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THE AD AND CSC COMMUNITIES

- The **computational graph** (Bauer, 1974) opens the door for many combinatorial problems in AD, and yet researchers outside the AD community are not familiar with them.
- The AD research community has organized since 1991, and the **combinatorial scientific computing (CSC)** community has organized since 2004.
- Hovland, Naumann, Walther, Utke, Lyons, and others from AD have been interacting with the CSC community since its inception.
- This talk tells a part of this story, and looks forward to even more fruitful interactions.



- An Example of the Partnership:
The CSCAPES Institute
- Describing the Partnership:
Combinatorics (Coloring) in AD
- Strengthening the Partnership:
A Way Forward



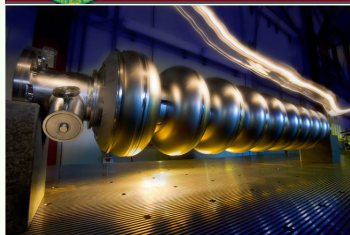
1 THE CSCAPES INSTITUTE

2 COMBINATORICS IN AD

- Coloring Models
- Sequential algorithms
- Case studies
- Parallel algorithms
- Summary

3 THE WAY FORWARD

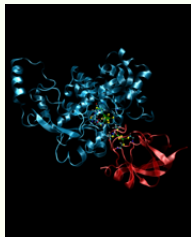




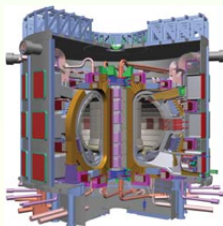
Accelerator Design



RF Gun, SLAC



Green Energy



ITER



SCIENTIFIC DISCOVERY THRU' ADVANCED COMPUTING II

- Office of Science, U.S. Dept. of Energy, 2006-2011
- Sciences
 - Fusion, Accelerator Design, Cosmology, Quantum Chromodynamics, Climate, Groundwater, Materials, Life Sciences
- Enabling Technologies
 - Applied Mathematics, Computer Science, Visualization, Data Management
- ET teams collaborating with Science Application teams to harness petascale computing in simulations
- \$ 90 Million per year, 30+ projects, 70+ institutions
- Involve computation at the beginning of a scientific research project, rather than as an afterthought.



CSCAPES INSTITUTE: PERSONNEL

- Purdue (Old Dominion): Alex Pothén, [Assefaw Gebremedhin](#), Florin Dobrian, Mahantesh Halappanavar, Brandon Hill, Min Huang, Duc Nguyen
- Sandia: [Erik Boman](#), Karen Devine, Bruce Hendrickson, Cederic Chevalier, Michael Wolfe
- Argonne: [Paul Hovland](#), Boyana Norris, Jean Utke, Ilya Safro, Andrew Lyons
- Ohio State: [Umit Catalyurek](#), Doruk Bozdog
- Colorado State: [Michelle Mills Strout](#)
- Other Collaborators: Andrea Walther, Uwe Naumann, Fredrik Manne, Yoshi Kawajiri, Larry Biegler



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- Computational graph (Bauer, 1974)
- Graph transformations
 - Vertex, edge, face elimination (Griewank, Forth, Pryce, Tadjouddine, Naumann, Utke...)
- Graph coloring (Coleman, More, Steihaug, Hossain, Verma, ...)
- Checkpoint placement in reverse mode (Griewank, Walther, ...)
- ...



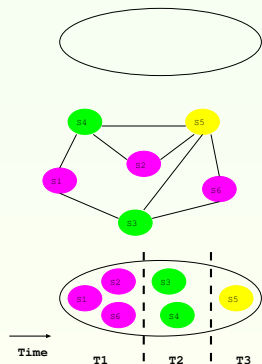
COLORING IN PARALLEL PROCESSING

- A **distance-1 coloring** of $G = (V, E)$ is
 - a mapping $\phi: V \rightarrow \{1, 2, \dots, q\}$ s.t.
 $\phi(u) \neq \phi(v)$ whenever $(u, v) \in E$
 - a partitioning of V into q **independent sets**

The objective is to **minimize** q

- Distance-1 coloring is used to **discover concurrency** in parallel scientific computing.
Examples:

- iterative methods for sparse linear systems (Jones & Plassmann, 94)
- adaptive mesh refinement
- preconditioners (Saad, 96; Hysom & Pothen, 01)
- eigenvalue computation (Manne, 98)
- sparse tiling (Strout et al, 02)



Procedure SPARSECOMPUTE($F : R^n \rightarrow R^m$)

- S1.** Determine the **sparsity structure** of the derivative (first or second) matrix $A \in R^{m \times n}$ of the function F
- S2.** Obtain a **seed** matrix $S \in \{0, 1\}^{n \times q}$ with the smallest q
- S3.** Compute the numerical values of the entries of the **compressed** matrix $B = AS \in R^{m \times q}$
- S4.** **Recover** the numerical values of the entries of A from B

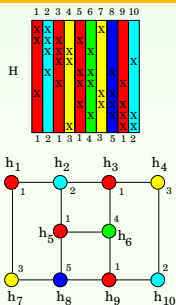
The seed matrix S **partitions** the columns of A :

$$s_{jk} = \begin{cases} 1 & \text{iff column } a_j \text{ belongs to group } k, \\ 0 & \text{otherwise.} \end{cases}$$

It is obtained using an appropriate **coloring** on the graph of A .



AN ACCURATE MODEL FOR DIRECT HESSIAN COMPUTATION

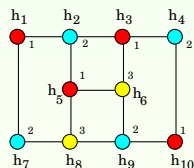
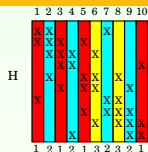


$$\begin{pmatrix} & h_{11} & h_{12} & h_{17} & 0 & 0 \\ h_{21} + h_{23} + h_{25} & h_{22} & 0 & 0 & 0 & 0 \\ & h_{33} & h_{32} & h_{34} & h_{36} & 0 \\ & h_{43} & h_{4,10} & h_{44} & 0 & 0 \\ & h_{55} & h_{52} & 0 & h_{56} & h_{58} \\ h_{63} + h_{65} + h_{69} & 0 & 0 & 0 & h_{66} & 0 \\ & h_{71} & 0 & h_{77} & 0 & h_{78} \\ & h_{85} + h_{89} & 0 & 0 & 0 & h_{88} \\ & h_{99} & h_{9,10} & 0 & h_{96} & h_{98} \\ h_{10,9} & h_{10,10} & h_{10,4} & 0 & 0 & 0 \end{pmatrix}$$

- **Symmetrically orthogonal partition:** whenever $h_{ij} \neq 0$
 - h_j only column in a group with nonzero at row i or
 - h_i only column in a group with nonzero at row j
- **Star coloring:** a vertex coloring ϕ of $G_a(H)$ s.t.
 - ϕ is a distance-1 coloring and
 - every path on 4 vertices (P_4) uses at least 3 colors
- SymOP equivalent to star coloring (Coleman and Moré, 84)



AN ACCURATE MODEL FOR HESSIAN COMPUTATION VIA SUBSTITUTION



$$\begin{pmatrix} h_{11} & h_{12} + h_{17} & 0 \\ h_{21} + h_{23} + h_{25} & h_{22} & 0 \\ h_{33} & h_{32} + h_{34} & h_{36} \\ h_{43} + h_{4,10} & h_{44} & 0 \\ h_{55} & h_{52} & h_{56} + h_{58} \\ h_{63} + h_{65} & h_{69} & h_{66} \\ h_{71} & h_{77} & h_{78} \\ h_{85} & h_{87} + h_{89} & h_{88} \\ h_{9,10} & h_{99} & h_{96} + h_{98} \\ h_{10,10} & h_{10,4} + h_{10,9} & 0 \end{pmatrix}$$

- **Substitutable partition:** whenever $h_{ij} \neq 0$
 - h_j in a group where all nonzeros in row i are ordered before h_{ij} or
 - h_i in a group where all nonzeros in row j are ordered before h_{ij}
- **Acyclic coloring:** a vertex coloring ϕ of $G_a(H)$ s.t.
 - ϕ is a distance-1 coloring and
 - every cycle uses at least 3 colors
- Substitutable partition equivalent to acyclic coloring (Coleman and Cai, 86)



General sparsity pattern:

	unidirectional partition	bidirectional partition	
Jacobian	distance-2 coloring	star bicoloring	Direct
Hessian	star coloring (restricted star coloring)	NA	Direct
Jacobian	NA	acyclic bicoloring	Substitution
Hessian	acyclic coloring (triangular coloring)	NA	Substitution

$$\begin{array}{ll} \text{Nonsym } A & G_b(A) = (V_1, V_2, E) \\ \text{Sym } A & G(A) = (V, E) \end{array}$$

Regular sparsity pattern (discretization of structured grids):

- Formula-based coloring (Goldfarb and Toint, 1984)
- Hierarchical coloring (Hovland, 2007)



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COMPLEXITY AND ALGORITHMS

- Distance- k , star, and acyclic coloring are NP-hard (they are also hard to approximate)

- A greedy heuristic usually gives a good solution

GREEDY($G = (V, E)$)

Let v_1, v_2, \dots, v_n be an **ordering** of V

for $i = 1$ to n **do**

 determine colors **forbidden** to v_i

 assign v_i the **smallest** permissible color

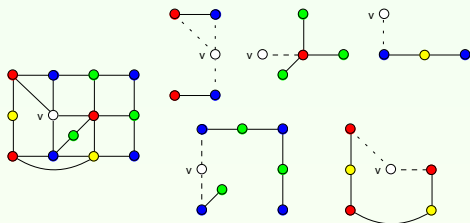
end-for

- For distance- k coloring, **GREEDY** can be implemented to run in $O(n\bar{d}_k)$ time, where \bar{d}_k is the average degree- k
- We have developed $O(n\bar{d}_2)$ -time heuristic algorithms for star and acyclic coloring

Key idea: exploit the structure of **two-colored induced subgraphs**



A NEW ACYCLIC COLORING HEURISTIC ALGORITHM



Algorithm (Input: $G = (V, E)$):

for each $v \in V$

① Choose color for v

- forbid colors used by neighbors $N(v)$ of v
- forbid colors leading to two-colored cycles
 - \forall tree T incident on v , if v adj to ≥ 2 vertices of *same* color, forbid the other color in T

② Update collection of two-colored trees (merge if necessary)

Time: $O(|V|\bar{d}_2 \cdot \alpha)$ **Space:** $O(|E|)$



NUMBER OF COLORS

Theorem: For every **chordal** graph $G = (V, E)$

$$\omega(G) = \chi_1(G) = \chi_a(G) \leq \beta(G) + 1$$

$$\chi_s(G) \leq \chi_2(G) = \omega(G^2) \leq \min \{2\beta(G) + 1, |V|\}$$

All inequalities become equalities when G is a **band** graph.

Experimental results:

$\rho, \bar{\rho}$	10, 10.98		20, 20.99	
	star	acyclic	star	acyclic
banded	11	6	21	11
random	21 – 24	9 – 11	50 – 56	18 – 19

Observed results for banded matrices are *optimal*.



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S1: Sparsity Detection

S2: Coloring

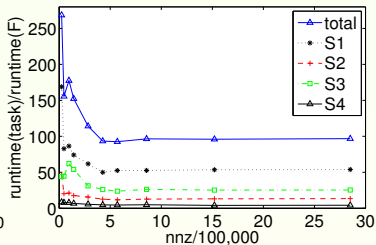
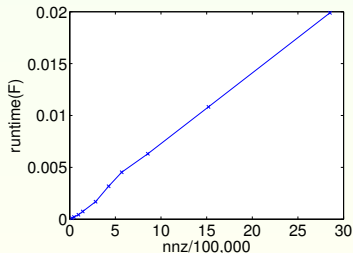
S3: Computation of Compressed Matrices

S4: Recovery of derivative matrix elements



EXPERIMENTS USING ADOL-C

- Efficacy of the four-step scheme tested in two case studies
 - ① Jacobian computation in a **Simulated Moving Bed** process (chromatography) **Walther AD08 talk**
 - ② Hessian computation in an optimal **electric power flow** problem
- Experiments showed
 - technique enabled cheap Jacobian/Hessian computation where dense computation is infeasible
 - observed results for each step matched analytical results



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PARALLELIZING GREEDY COLORING

- Desired task: parallelize **GREEDY** such that
 - speedup is $\Theta(p)$
 - number of colors used is roughly same as in serial
- A difficult task since **GREEDY** is inherently sequential
- For D1 coloring, several approaches based on Luby's parallel algorithm for **maximal independent set** exist
- Some drawbacks:
 - no actual parallel implementation
 - many more colors than a serial implementation
 - poor parallel speedup on unstructured graphs



(RELAXED) PARTITIONING APPLIED TO GREEDY COLORING

Basic features of the algorithm:

- exploits features of data distribution
 - distinguishes between **interior** and **boundary** vertices
- proceeds in **rounds**, each having **two phases**:
 - **tentative coloring**
 - **conflict detection**
- tentative coloring phase organized in **supersteps**
 - each processor communicates **only after** coloring a subset of its assigned vertices using currently available information (infrequent, coarse-grain communication)
- **randomization** used in resolving conflicts



A FRAMEWORK FOR PARALLEL DISTANCE-1 COLORING

FRAMEWORK($G = (V, E), s$)

Partition V into V_1, V_2, \dots, V_p using a **graph partitioner**

(Processor P_i owns (and colors) V_i , and stores edges E_i incident on V_i .)

On each processor $P_i, i \in I = \{1, \dots, p\}$

while uncolored vertices remain **do** (rounds)

Partition uncolored vertices into subsets of size s

(supersteps for tentative coloring)

for each superstep **do**

Tentatively color vertices in the superstep

Send colors of boundary vtxs to relevant processors

Receive color information from relevant processors

Wait until all incoming messages are received

Detect conflicts for boundary vertices

Uncolor incorrectly colored boundary vertices

SPECIALIZATIONS OF FRAMEWORK

FRAMEWORK can be specialized along several axes:

- 1 **Color selection strategies:**
 - First Fit: search for smallest color starts at 1 on each processor
 - Staggered FF: search for smallest color starts from different “bases”
- 2 **Coloring order:**
 - interior vertices can be colored **before**, **after**, or **interleaved with** boundary vertices
- 3 **Local vertex ordering:**
 - vertices on each processor can be ordered using various **degree-based** techniques
- 4 **Supersteps:**
 - can be run **synchronously** or **asynchronously**
- 5 **Inter-processor communication:**
 - can be **customized** or **broadcast-based**



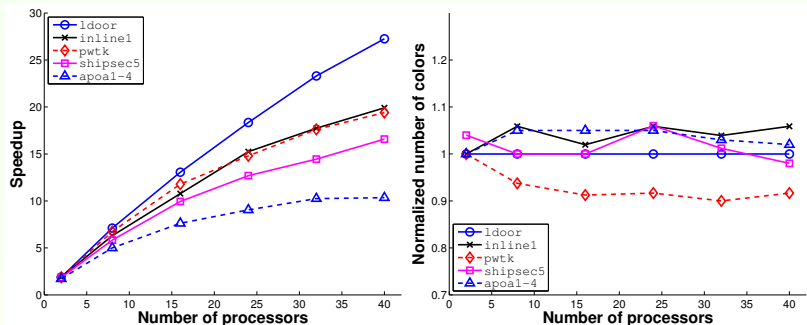
LESSONS LEARNED FROM EXPERIMENTS

Good parameter configuration for large-size (millions of edges) graphs:

- moderately unstructured graphs (e.g. a typical application graph):
 - 1 a superstep size s in the order of 1000
 - 2 asynchronous supersteps
 - 3 a coloring order in which interior vertices appear either strictly before or strictly after boundary vertices
 - 4 First Fit color choice strategy
 - 5 customized inter-processor communication
- highly unstructured (e.g. random) graphs:
 - s in the order of 100
 - items 2 to 4 same as for moderately unstructured graphs
 - broadcast-based communication



A SAMPLE EXPERIMENTAL RESULT: STRONG SCALABILITY



Algorithm FBAC on Itanium 2 cluster.



SUMMARY

- **Current accomplishments:**
 - Designed and implemented new sequential algorithms for distance- k , star, acyclic, and other coloring problems.
 - C++ implementations and ordering functions assembled in a package called **ColPack**.
 - Integrated parts of **ColPack** with the AD tool ADOL-C.
 - Developed parallel algorithms for distance-1, distance-2, and restricted star coloring, available through the **Zoltan** package.
- **Planned activities:**
 - Integrate coloring software with tools in OpenAD.
 - Develop parallel coloring algorithms for tera- and peta-scale computation.
 - Collaborate with chemical engineers to solve coupled PDEs with cyclic boundary conditions from liquid chromatographic and gaseous separations.

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MORE ABOUT CSCAPES, SciDAC

- www.cscapes.org, www.scidac.org
- CSCAPES article in SciDAC Review, Fall 2007, pp. 26–35.
- CSCAPES Institute seminars using remote conferencing systems
- SIAM Workshop on CSC 2009, Oct. 29-31, 2009 (Monterey Beach CA)
- Oberwolfach Workshop on CSC, Feb. 2009
- Annual SciDAC conferences, SciDAC Review (online magazine)
- **DEISA**: Distributed European Infrastructure for Supercomputer Applications.



MORE ABOUT AD AND CSC COMMUNITIES

- Need to increase participation in our communities. Both have steep learning curves as price of admission. Planned survey by Gebremdhin and Naumann on Combinatorics in AD.
- Strengthen existing collaborations in CSCAPES Institute and elsewhere.
- Develop robust AD software efficient for many core architectures.
- Collaborate with researchers working on tera- and peta-scale simulations: intensive computations, large data sets, deep memory hierarchies, and non-uniform communication costs: opportunities for combinatorial algorithms.



FURTHER READING



Hendrickson and Pothen,
CSC: The enabling power of discrete algorithms in computational science.
LNCS, 4395, pp. 260-280, 2007.



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SIAM J. Sci. Comput. 29:1042–1072, 2007.



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Exploiting Sparsity in Jacobian Computation via Coloring and AD:
A Case Study in a Simulated Moving Bed Process.
Fifth International Conference on AD, Bonn, Germany, Aug 2008, to appear. 12 pp.



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INFORMS Journal on Computing, to appear. 30 pp.



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J. Parallel Distrib. Comput. 68(4):515–535, 2008.