

Toward Advances in 3 Areas
Dept. 1411 Review, May 2006

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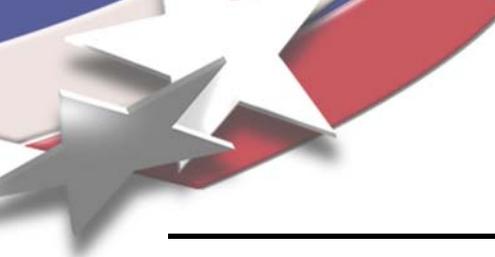
SAND2006-3620P

Sandia is a multiprogram laboratory operated by Sandia Corporation, a Lockheed Martin Company, for the United States Department of Energy's National Nuclear Security Administration under contract DE-AC04-94AL85000.



Outline

- 1. Background**
- 2. Automatic Differentiation**
- 3. Global Optimization**
- 4. AMPL**
- 5. Service**



Background

- **Interests include**
 - **Scientific computing in general**
 - **Smooth optimization in particular**
- **Good experience using AD with AMPL**
 - **Gradient, Hessian computations on AMPL expression graphs**
- **Early experience bounding nonlin. eqn sol'ns**
- **Now at Sandia → chance to work on large-scale computations.**
 - **This year, focused on three disparate topics.**



Automatic Differentiation CSRF

- **Work with Eric Phipps (1416); dm_g = PI**
 - With helpful comments from Roscoe Bartlett, Andy Salinger, etc.
- **Goal: promote better use of computing resources via better algorithms for derivatives.**
- **Work this past year includes**
 - More work with Charon (Eric Phipps)
 - Improving, templating RAD (dm_g)
 - Paper for ICCS06 (Bartlett, Gay, Phipps)
 - Eric will present in Reading, UK, late May



Need for derivatives

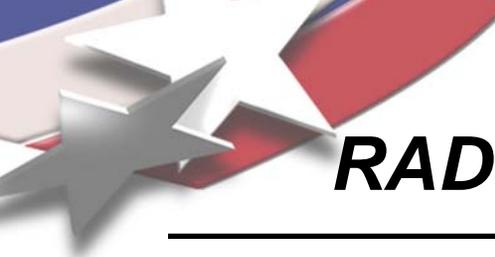
- **Nonlinear equations: solving $f(\mathbf{x}) = 0$**
 - Newton's method: $\mathbf{x}^{k+1} = \mathbf{x}^k - \mathbf{f}'(\mathbf{x}^k)^{-1} \mathbf{f}(\mathbf{x}^k)$
- **ODEs, PDEs, DAEs**
- **Optimization**
 - Minimize $f(\mathbf{x})$ s.t. $\mathbf{L} \cdot \mathbf{c}(\mathbf{x}) \cdot \mathbf{u}$
 - Find β to minimize $\sum_i [y_i - M(\mathbf{x}_i, \beta)]^2$
- **Sensitivities (first order)**
 - How do small changes in \mathbf{x} affect $f(\mathbf{x})$?
 - Useful in Verification and Validation

What if we could compute as though we had several Red Storms?



Action on a Mesh

- In large-scale computation, values of interest are often computed on elements of a mesh.
- Often a small number of mesh-element functions.
- Absent time stepping,
Visiting each element once and
computing its contribution to f , rf , r^2f and
Manually assembling the overall f
may be much more efficient than straightforward
backward AD.



RAD*: Specialized overloading for *rf

- “Active” variables of type ADvar
 - Simply assigned values initially
 - Can be overwritten
 - Each operation stores values, partials in memory

- **Invoke**

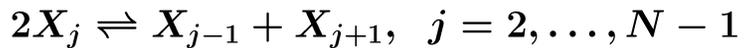
```
ADcontext::Gradcomp();
```

when gradient desired.

- Causes reverse sweep and reclamation of memory
 - For ADvar *v*,
 - *v.val()* = current value
 - *v.adj()* = adjoint w.r.t. last computed ADvar value
- **Block memory allocation for efficiency.**

Efficiency of AD in Charon

Set of N hypothetical chemical species:



Steady-state mass transfer equations:

$$\nabla^2 Y_j + \mathbf{u} \cdot \nabla Y_j = \dot{\omega}_j, \quad j = 1, \dots, N - 1$$

$$\sum_{j=1}^N Y_j = 1$$

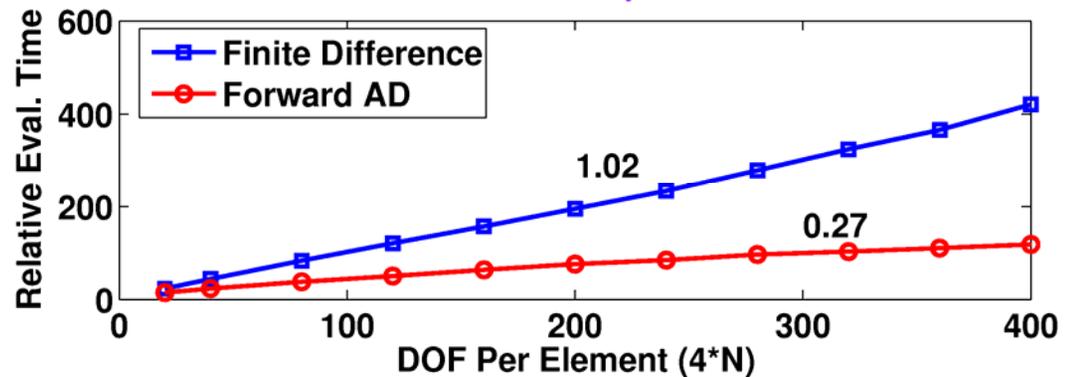
Automatic Differentiation

- ✓ Faster than FD
- ✓ Better scalability in number of PDEs
- ✓ Analytic derivative

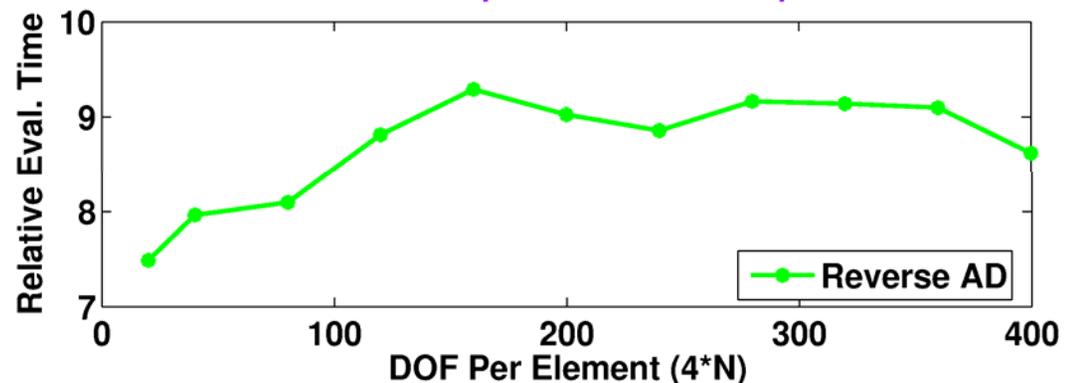
Results published, presented at ICCS 2006, Reading, UK

Efficiency of the element-level derivative computation

Jacobian Computation

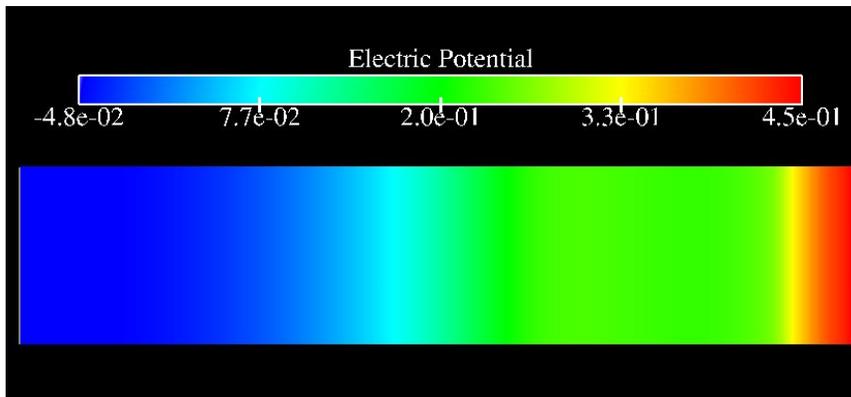


Jacobian-Transpose Product Computation



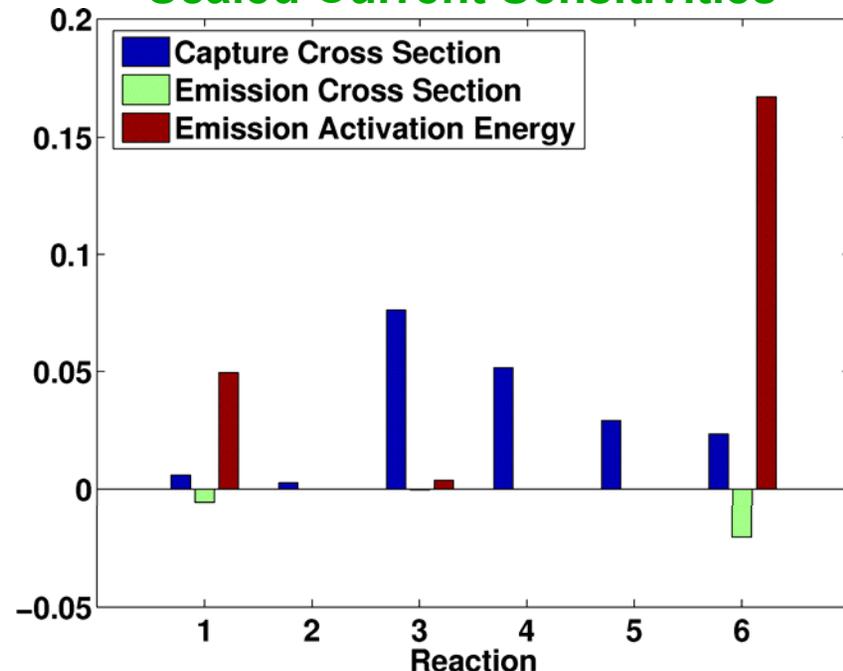
Steady-State Sensitivity Analysis of 1n5614 Diode -- 0.5V Bias

4 defect species, 2 currents, 18 parameters, 126k unknowns, 2 procs



$$\frac{\Delta I}{I} \approx \left(\frac{p}{I} \frac{\partial I}{\partial p} \right) \frac{\Delta p}{p}$$

Scaled Current Sensitivities



#	Reaction	Cross Section	Emission Activation Energy
1	$VV^- + e^- \leftrightarrow VV^{--}$	$3.0e-16$	0.23
2	$VV^{--} + h^+ \leftrightarrow VV^-$	$3.0e-14$	0.89
3	$VV^0 + e^- \leftrightarrow VV^-$	$3.0e-15$	0.41
4	$VV^- + h^+ \leftrightarrow VV^0$	$3.0e-14$	0.71
5	$VV^+ + e^- \leftrightarrow VV^0$	$3.0e-14$	0.91
6	$VV^0 + h^+ \leftrightarrow VV^+$	$3.0e-15$	0.21

Adjoint sensitivities > 9 times faster than direct!



RAD improvements

- **More templating (for use with various types)**
 - Easier use with constants (integer, floating-point)
 - Interacts successfully with FAD
 - Hessian-vector products
- **Means of interacting with 3rd-party functions**
 - E.g., in Charon, Chemkin processed by Adifor
- **Allow multiple reverse sweeps for several outputs**
- **More debugging facilities**
 - `RAD_DEBUG_BLOCKKEEP=n` (puts NaNs into n old memory blocks)
 - `RAD_AUTO_AD_Const` treats unfreed ADvar's as constants



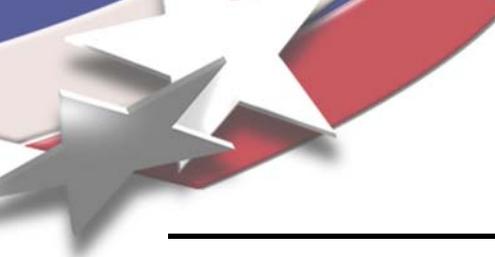
AD outreach at Sandia

- **Presentation at Sandia Software Developer's Conference (dmg, May 2006)**
- **Interest in more widespread AD introduction**

Working on...

- **Web site for *Sacado***
- **Examples**

Goal: facilitate code owner's use of AD



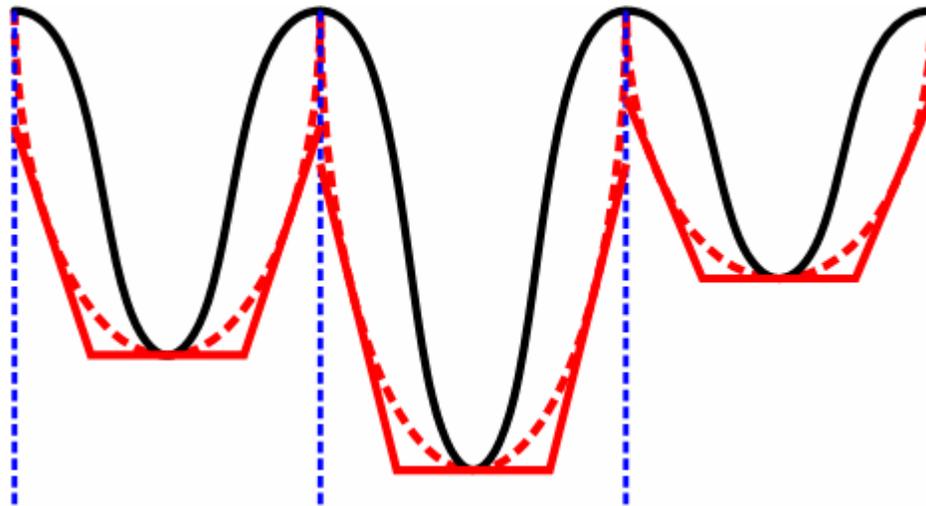
AD Summary

- **Programmatic issue: better algs for derivatives**
- **Activities: develop *Sacado* = Sandia AD tools**
- **Technical:**
 - New RAD package for reverse AD (dmg)
 - Improved FAD package for fwd AD (etphipp)
- **Results: good experience in Charon**
- **Impact:**
 - More efficient computation
 - Banish confusion from truncation error
 - Simplify maintenance — save programmer time

Rigorous Global Optimization

PI: Bill Hart

Application example: NW safing





FY06 Rigorous Global Optimization

- **Bill Hart, PI**
- **Dave Gay**
- **Hosted two summer students in 2005**
 - **Lianjun Jiang (U of CO)**
 - interval slopes, other Coconut facilities
 - **Pradeep Polisetty (U of S. Carolina)**
 - convex underestimators from expression graph
- **Adding “gnlp” to Acro**
- **Rewrote “ampltodag”**
 - Indication of integer variables



Global optimization summary

- **Programmatic issue:** minimize or maximize a compactly constrained objective
- **Activities:** integration of rigorous bounding and fathoming tests (exclusion, sufficiency) into PICO parallel branch-and-bound framework
- **Technical:** developing *gnlp*
- **Results:** working toward first *gnlp* release
- **Impact:**
 - NW safing
 - Enable asking, “How well could we do?”
 - Decision support



AMPL extensions

FY06 MICS support for extensions to stochastic programming (decision support, ISE)

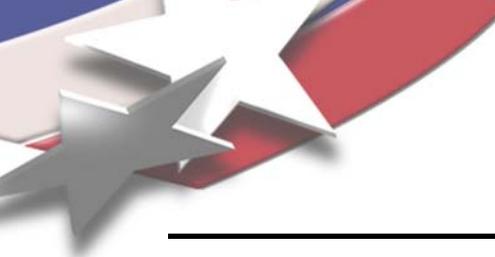
AMPL: a language for *mathematical programming*:

minimize $f(x)$

s.t. $L \cdot c(x) \cdot u$

With $x \in \mathbb{R}^n$ and $c: \mathbb{R}^n \rightarrow \mathbb{R}^m$

Given algebraically and some x_i discrete.



AMPL goals/features

- **Easy transcription from math (*avoid mistakes*)**
- **Explicit indexing (*no hidden magic*)**
- **Declare before use (*one-pass reading*)**
- **Separate model, data, commands (*orthogonality*)**
- **Separate solvers (*open solver interface*)**
- **Update entities as needed (*lazy evaluation*)**
- **Builtin math. prog. Stuff (*presolve, red. costs, ...*)**
- **Design for large scale nonlinear (*sparsity, algs.*)**
- **Allow extensions via shared libraries**
 - Imported functions, table handlers



Stochastic programming examples

Recourse problems:

Minimize $E(\text{today's decision} + \text{tomorrow's corrections})$

Probabilistic constraints:

$\text{Prob}(\text{failure}) \leq \varepsilon$

$\text{Prob}(\text{success}) \geq 1 - \varepsilon$



AMPL extension: random variables

```
var v random;  
# ...  
s. t. Constr1: ... >= v;  
s. t. Constr2: c(v) == 0;  
# ...  
let v := Normal 0 1 () + Uniform(-1, 2);  
# or  
let v := 4.2;
```

Let the solver do the sampling.



Implementation goals

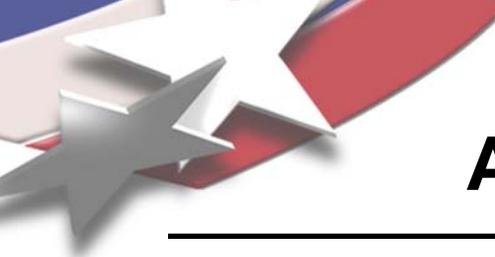
Interface to Gassmann's MSLip, SMPS.

Interface to Gerd Infanger's importance-sampling solver.



AMPL support at SNL

- Overview talk to 1400 last Oct.
 - Overview talk to JWAC last Nov.
 - Interactions with 6221 re table handler.
 - Binaries for AMPL, several solvers in dropzone
 - Re pcx, Craig Lawton (6642) said “Wow, this is going to help me quite a bit.”
 - Recent ISE model change
 - s.t. $C\{(i,j,k) \text{ in } A, (i,j,p) \text{ in } A, (p,k) \text{ in Parent}\}: \dots$
 - to
 - s.t. $C\{(p,k) \text{ in Parent}, (i,j,k) \text{ in } A: (i,j,p) \text{ in } A\}: \dots$
- reduced instantiation time from 10 hrs to 40 sec.



AMPL extensions summary

- **Programmatic issue: stochastic decision support**
- **Activities: stochastic-programming extensions to AMPL**
- **Thesis: let the solver do the sampling**
- **Results: work in progress**
- **Impact: better design, decision making**
 - **Another tool for handling uncertainty**



Service

- **Math. Programming Society Treasurer**
- **AMPL support**

R&D and international support at home

Recently:

indicator constraints in CPLEX

(special case of constraint programming)

converting logical constr's to algebraic in presolve

- **Refereeing, papers for**
 - **SCAN04 proceedings**
 - **ACM Trans. Math. Software**

Automatic Differentiation (AD) CSRF Project Funded for FY05-07

- Acquired funding for a new CSRF project on AD with David Gay (PI, 9211)
- Develop AD tools for Sandia DP applications
 - Forward: Extend and “robustify” TFAD
 - Reverse: Investigate efficiency, scalability of operator overloading approaches
 - Taylor: Study Taylor polynomial based algorithms for time integration, uncertainty quantification
- Goals:
 - Make AD a commonplace technology at Sandia
 - Provide derivative technology to foster new algorithm development

Finite element residual equations

$$f(x) = 0$$

$$\frac{\partial f}{\partial x} V$$

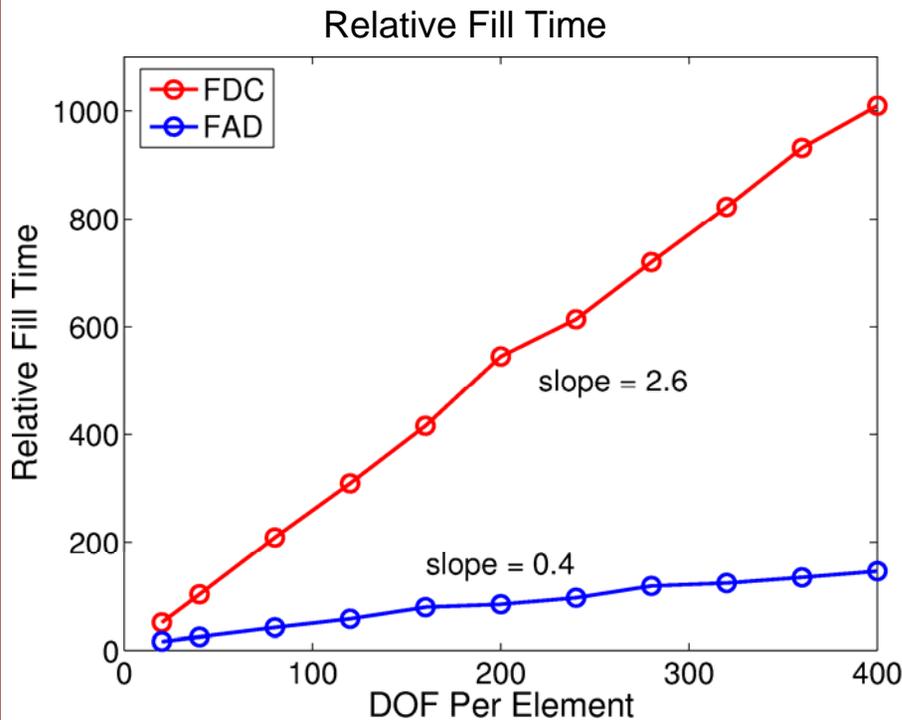
$$\left(\frac{\partial f}{\partial x}\right)^T V$$

$$x(t) = \sum_{k=0}^d x_k t^k$$
$$f(x(t)) = \sum_{k=0}^d f_k t^k + O(t^{d+1})$$

“Analytic derivatives are critical for the success of the Charon/QASPR project.” – Rob Hoekstra, Charon Project Lead

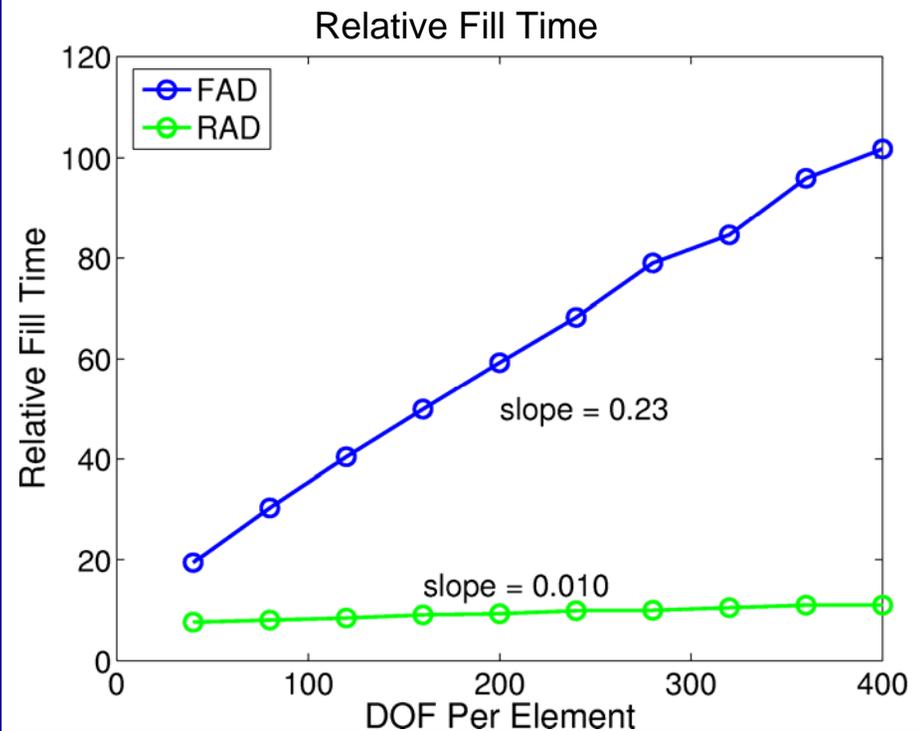
Forward and Reverse Mode AD in Charon

Forward Mode Jacobian Using FAD Versus Finite Differencing w/Coloring



FAD has become the standard Jacobian strategy in Charon

Reverse Mode Jacobian-Transpose Product Versus Forward Mode



Enabling technology for large-scale sensitivity analysis & optimization