

Enabling Predictive Simulation Through Embedded Sensitivity Analysis and Uncertainty Quantification

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CIS External Panel Review
April 13-16, 2008

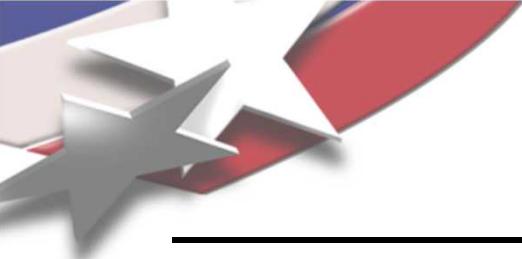
Tight Coupling Through Automatic Differentiation Enables Predictive Simulation

- Transformation to predictive science
 - V&V, UQ, QMU
- Single-point forward simulations are insufficient
 - Forward & adjoint sensitivities
 - Error estimates
 - Probability distributions
 - Intervals

MATURITY ELEMENT	Maturity Level 0 Low Consequence, Minimal M&S Impact, e.g., Scoping Studies	Maturity Level 1 Moderate Consequence, Some M&S Impact, e.g., Design Support	Maturity Level 2 High-Consequence, High M&S Impact, e.g., Qualification Support	Maturity Level 3 High-Consequence, Decision Making Based on M&S, e.g., Qualification or Certification
Representation and Geometric Fidelity What features are neglected because of simplifications or stylizations?	<ul style="list-style-type: none"> • Judgment only • Little or no representational or geometric fidelity for the system and boundary conditions (BCs) 	<ul style="list-style-type: none"> • Significant simplification or stylization of the system and BCs • Geometry or representation of major components is defined 	<ul style="list-style-type: none"> • Limited simplification or stylization of major components and BCs • Geometry or representation is well defined for major components and some minor components • Some peer review conducted 	<ul style="list-style-type: none"> • Essentially no simplification or stylization of components in the system and BCs • Geometry or representation of all components is at the detail of "as built"; e.g., gaps, material interfaces, fasteners • Independent peer review conducted
Physics and Material Model Fidelity How fundamental are the physics and material models and what is the level of model calibration?	<ul style="list-style-type: none"> • Judgment only • Model forms are either unknown or fully empirical • Few, if any, physics-informed models • No coupling of models 	<ul style="list-style-type: none"> • Some models are physics-based and are calibrated using data from related systems • Few, if any, physics-informed models • Some model or ad hoc coupling of models 	<ul style="list-style-type: none"> • Physics-based models for all important processes • Significant calibration needed using separate-effects tests (SETs) and integral-effects tests (IETs) • One-way coupling of models • Some peer review conducted 	<ul style="list-style-type: none"> • All models are physics based • Minimal need for calibration using SETs and IETs • Sound physical basis for extrapolation and coupling of models • Full, two-way coupling of models • Independent peer review conducted
Code Verification Are algorithm deficiencies, software errors, and poor SOE practices corrupting the simulation results?	<ul style="list-style-type: none"> • Judgment only • Minimal testing of any software elements • Little or no SOE procedures specified or followed 	<ul style="list-style-type: none"> • Code is managed by SQE procedures • Unit and regression testing conducted • Some comparisons made with benchmarks 	<ul style="list-style-type: none"> • Some algorithms are tested to determine the observed order of numerical convergence • Some features & capabilities (F&Cs) are tested with benchmark solutions • Some peer review conducted 	<ul style="list-style-type: none"> • All important algorithms are tested to determine the observed order of numerical convergence • All important F&Cs are tested with rigorous benchmark solutions • Independent peer review conducted
Solution Verification Are numerical solution errors and human procedural errors corrupting the simulation results?	<ul style="list-style-type: none"> • Judgment only • Numerical errors have unknown or large effect on simulation results 	<ul style="list-style-type: none"> • Numerical effects on relevant SRQs are qualitatively estimated • Input/output (I/O) verified only by the analysts 	<ul style="list-style-type: none"> • Numerical effects are quantitatively estimated to be small on some SRQs • I/O independently verified • Some peer review conducted 	<ul style="list-style-type: none"> • Numerical effects are determined to be small on all important SRQs • Important simulations are independently reproduced • Independent peer review conducted
Model Validation How carefully is the accuracy of the simulation and experimental results assessed at various tiers in a validation hierarchy?	<ul style="list-style-type: none"> • Judgment only • Few, if any, comparisons with measurements from similar systems or applications 	<ul style="list-style-type: none"> • Quantitative assessment of accuracy of SRQs not directly relevant to the application of interest • Large or unknown experimental uncertainties 	<ul style="list-style-type: none"> • Quantitative assessment of predictive accuracy for some key SRQs from IETs and SETs • Experimental uncertainties are well characterized for most SETs, but poorly characterized for IETs • Some peer review conducted 	<ul style="list-style-type: none"> • Quantitative assessment of predictive accuracy for all important SRQs from IETs and SETs at conditions/geometries directly relevant to the application • Experimental uncertainties are well characterized for all IETs and SETs • Independent peer review conducted
Uncertainty Quantification and Sensitivity Analysis How thoroughly are uncertainties and sensitivities characterized and propagated?	<ul style="list-style-type: none"> • Judgment only • Only deterministic analyses are conducted • Uncertainties and sensitivities are not addressed 	<ul style="list-style-type: none"> • Aleatory and epistemic (A&E) uncertainties propagated, but without distinction • Informal sensitivity studies conducted • Many strong UQ/SA assumptions made 	<ul style="list-style-type: none"> • A&E uncertainties segregated, propagated, and identified in SRQs • Quantitative sensitivity analyses conducted for most parameters • Numerical propagation errors are estimated and their effect known • Some strong assumptions made • Some peer review conducted 	<ul style="list-style-type: none"> • A&E uncertainties comprehensively treated and properly interpreted • Comprehensive SA conducted for parameters and models • Numerical propagation errors are demonstrated to be small • No significant UQ/SA assumptions made • Independent peer review conducted

Impacting CIS Transformation Mission

- Algorithms research for better efficiency, scalability, robustness
 - Tight algorithmic coupling to underlying physics
- Vehicle for incorporating these technologies into applications
 - Templating and (generalized) Automatic Differentiation (AD)



Crash Course in AD 101

- How does AD work
 - Derivatives at operation level known
 - Chain rule
- What does AD compute?
 - Forward mode derivatives (Jacobians, Jacobian-vector products)
 - Reverse mode derivatives (Gradients, Jacobian-transpose products)
 - High order univariate Taylor polynomials
- How is it implemented?
 - Source transformation (Fortran)
 - Operator overloading (C++)
- Multiple derivative components propagated simultaneously
 - Big cache benefit, potential for multi-core

AD Research Distinguished By Tools and Approach for Large-Scale Codes

- Many AD tools and research projects
 - ✗ Most geared towards Fortran (ADIFOR, OpenAD)
 - ✗ Most C++ tools are slow (ADOL-C)
 - ✗ Most applied in black-box fashion
- Sacado: Operator overloading AD tools for C++ applications
 - ✓ Multiple highly-optimized AD data types
 - ✓ Transform to template code & instantiate on Sacado AD types
 - ✓ Apply AD only at the “element level”
- Directly impacting QASPR through Charon
 - ✓ Analytic Jacobians and parameter derivatives
- This is the only successful, sustainable approach for large-scale C++ codes!



QuickTime™ and a
TIFF (Uncompressed) decompressor
are needed to see this picture.

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Efficient Sensitivity Analysis Requires Accurate Derivatives + Solver Integration

- Spatially discretized PDE:

$$f(\dot{x}, x, p, t) = 0$$

- Temporal discretization (Backward Euler):

$$f\left(\frac{x_{n+1} - x_n}{\Delta t}, x_{n+1}, p, t_{n+1}\right) = 0$$

- Forward sensitivity problem:

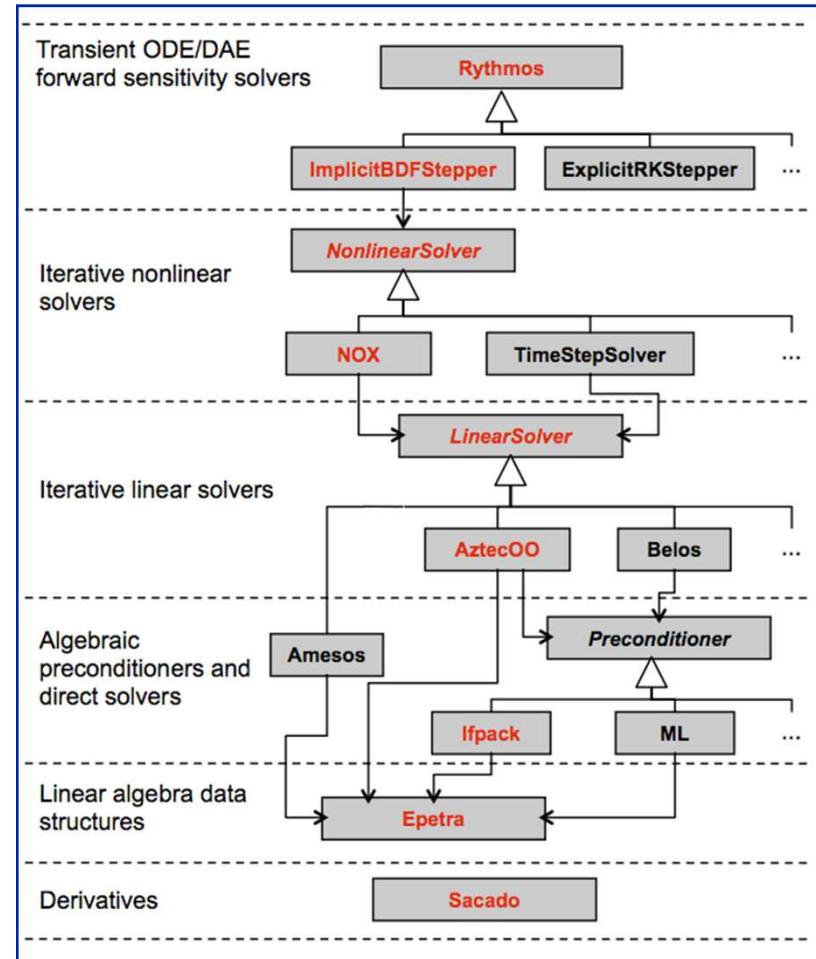
$$\frac{\partial f}{\partial \dot{x}}\left(\frac{\partial \dot{x}}{\partial p}\right) + \frac{\partial f}{\partial x}\left(\frac{\partial x}{\partial p}\right) + \frac{\partial f}{\partial p} = 0$$

$$\frac{1}{\Delta t} \frac{\partial f}{\partial \dot{x}}\left(\frac{\partial x_{n+1}}{\partial p} - \frac{\partial x_n}{\partial p}\right) + \frac{\partial f}{\partial x}\left(\frac{\partial x_{n+1}}{\partial p}\right) + \frac{\partial f}{\partial p} = 0$$

- Achieving transformation

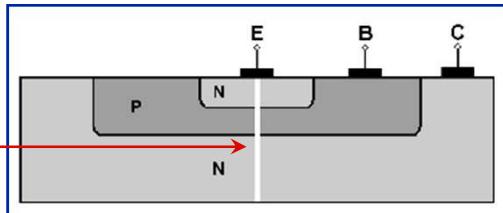
- Algorithm: Forward sensitivities
- Tight Coupling: Analytic derivatives
- Vehicle: AD + Templating

Vertical integration of Trilinos capabilities

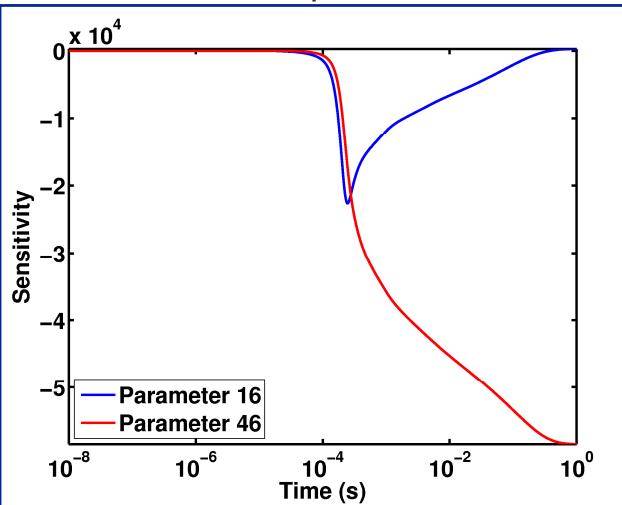


Capability Demonstrated on the QASPR Simple Prototype*

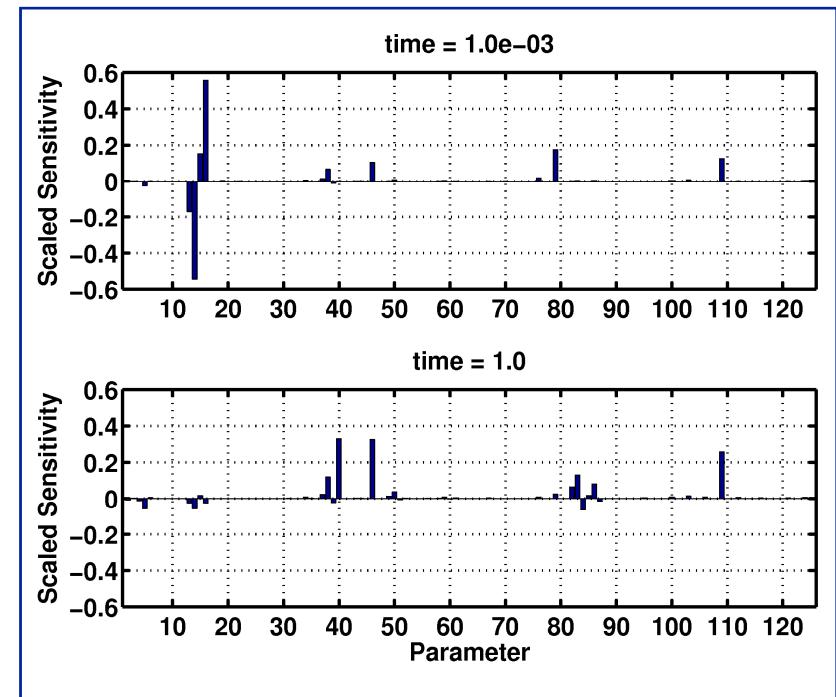
- Pseudo 1D Bipolar Junction Transistor (9x0.1 micron)
- Full defect physics
- 126 parameters



Sensitivities computed at all times



Sensitivities at early/late time show dominant physics



Comparison to standard FD approach:

- ✓ Sensitivities at all time points
- ✓ 14x faster
- ✓ More accurate
- ✓ More robust

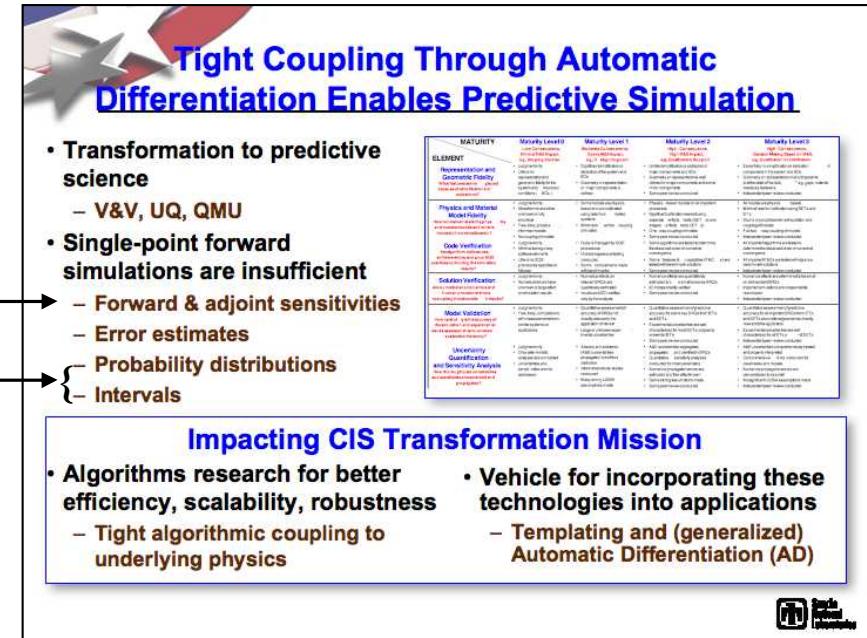
*ASC Level II Milestone (Bartlett et al)



Sandia
National
Laboratories

Impacting CIS Transformation to Predictive Science Through Templating and AD

- AD impact:
 - Forward sensitivity analysis (completed)
 - Adjoint sensitivity analysis (on going 1400-1500 collaboration)
- Templating provides deep interface to applications
- Impacting UQ:
 - Stochastic Galerkin methods (on going)
 - Taylor methods (near term)
 - Intervals (mid term)
 - Probability boxes (long term)



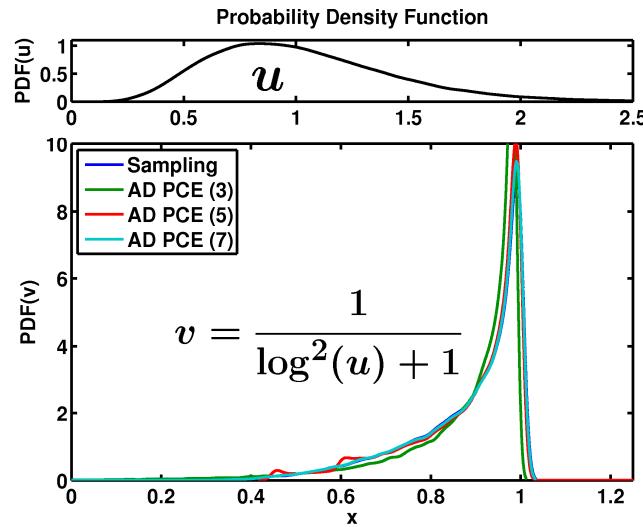
Templating Enables Automatic Embedded Uncertainty Propagation Research

Stochastic Galerkin method

$$f(x, \xi) = 0 \rightarrow \hat{x}(\xi) = \sum_{i=0}^P x_i \phi_i(\xi) \rightarrow f_i(x_0, \dots, x_P) \equiv \int f(\hat{x}(\xi), \xi) \phi_i(\xi) d\mu = 0$$

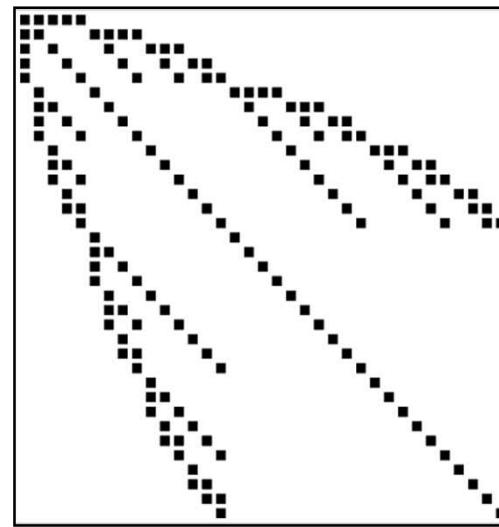
Embedded SG techniques using AD-like procedure (completed)

– Compute projection op-by-op

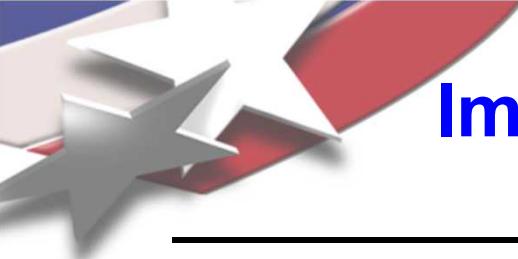


Trilinos interfaces for solving block SG linear systems (on going)

– Linear algebra research

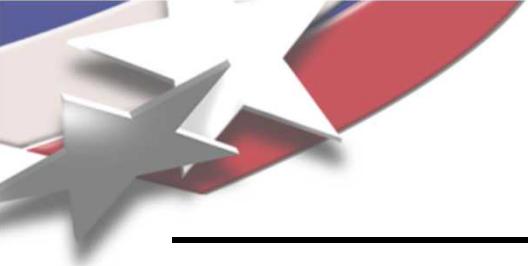


Encapsulated in new Trilinos package *Stokhos*
– Collaboration with Roger Ghanem (USC)



Impacting Sandia's Mission & External Community

- **Sandia mission impact**
 - **Rapid physics development**
 - Charon semiconductor device modeling for QASPR (ASC)
 - Charon chemical laser (Air Force WFO)
 - Charon extended MHD (ASCR)
 - SIERRA/Aria and Xyce (ASC)
 - **Charon transient sensitivity analysis for QASPR**
 - **Potential for automatic embedded UQ in all of these codes**
- **Internal recognition (including Algorithms Integration)**
 - **2 IPA awards, 2 ERA nominations, 1 ERA award**
- **External visibility**
 - **Collaboration with USC**
 - **3 publications**
 - **4 conference presentations**
 - **Program committee member for AD 2008**



Auxiliary Slides

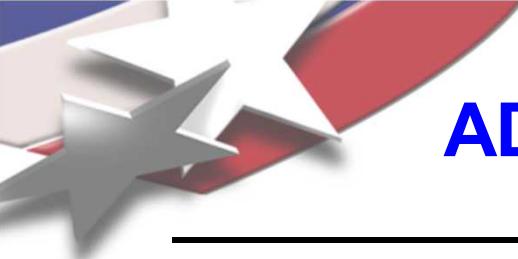
What is Automatic Differentiation (AD)?

- Technique to compute analytic derivatives without hand-coding the derivative computation
- How does it work -- freshman calculus
 - Computations are composition of simple operations (+, *, sin(), etc...) with known derivatives
 - Derivatives computed line-by-line, combined via chain rule
- Derivatives accurate as original computation
 - No finite-difference truncation errors
- Provides analytic derivatives without the time and effort of hand-coding them

$$y = \sin(e^x + x \log x), \quad x = 2$$

$x \leftarrow 2$	$\frac{dx}{dx} \leftarrow 1$
$t \leftarrow e^x$	$\frac{dt}{dx} \leftarrow t \frac{dx}{dx}$
$u \leftarrow \log x$	$\frac{du}{dx} \leftarrow \frac{1}{x} \frac{dx}{dx}$
$v \leftarrow xu$	$\frac{dv}{dx} \leftarrow u \frac{dx}{dx} + x \frac{du}{dx}$
$w \leftarrow t + v$	$\frac{dw}{dx} \leftarrow \frac{dt}{dx} + \frac{dv}{dx}$
$y \leftarrow \sin w$	$\frac{dy}{dx} \leftarrow \cos(w) \frac{dw}{dx}$

x	$\frac{d}{dx}$
2.000	1.000
7.389	7.389
0.301	0.500
0.602	1.301
7.991	8.690
0.991	-1.188



AD Takes Three Basic Forms

$$x \in \mathbf{R}^n, f : \mathbf{R}^n \rightarrow \mathbf{R}^m$$

- **Forward Mode:**

$$(x, V) \longrightarrow \left(f, \frac{\partial f}{\partial x} V \right)$$

- Propagate derivatives of intermediate variables w.r.t. independent variables forward
- Directional derivatives, tangent vectors, square Jacobians, $\partial f / \partial x$ when $m \geq n$

- **Reverse Mode:**

$$(x, W) \longrightarrow \left(f, W^T \frac{\partial f}{\partial x} \right)$$

- Propagate derivatives of dependent variables w.r.t. intermediate variables backwards
- Gradient of a scalar value function with complexity $\approx 4 \text{ ops}(f)$
- Gradients, Jacobian-transpose products (adjoints), $\partial f / \partial x$ when $n > m$

- **Taylor polynomial mode:**

$$x(t) = \sum_{k=0}^d x_k t^k \longrightarrow \sum_{k=0}^d f_k t^k = f(x(t)) + O(t^{d+1}), \quad f_k = \frac{1}{k!} \frac{d^k}{dt^k} f(x(t))$$

- **Basic modes combined for higher derivatives:**

$$\frac{\partial}{\partial x} \left(\frac{\partial f}{\partial x} V_1 \right) V_2, \quad W^T \frac{\partial^2 f}{\partial x^2} V, \quad \frac{\partial f_k}{\partial x_0}$$

Efficiency of AD in Charon

Set of N hypothetical chemical species:
 $2X_j \rightleftharpoons X_{j-1} + X_{j+1}, \quad j = 2, \dots, N - 1$

Steady-state mass transfer equations:

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

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

- **Forward mode AD**

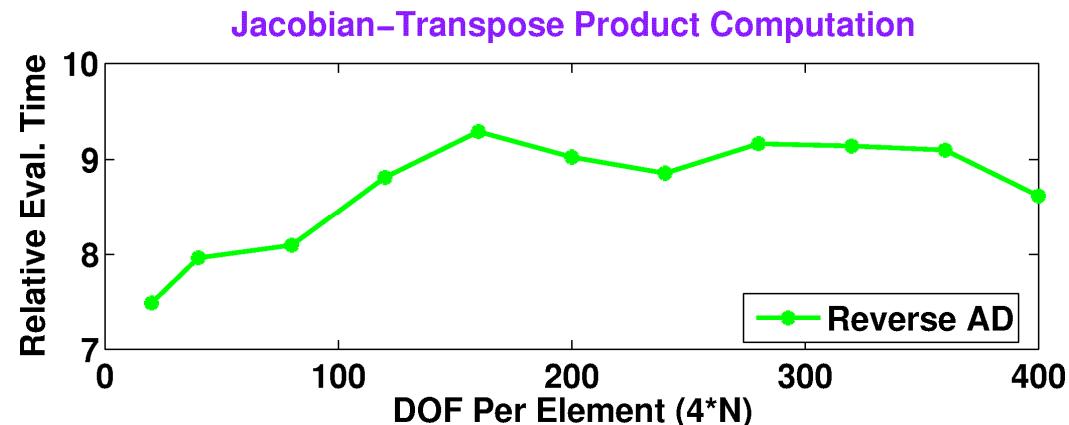
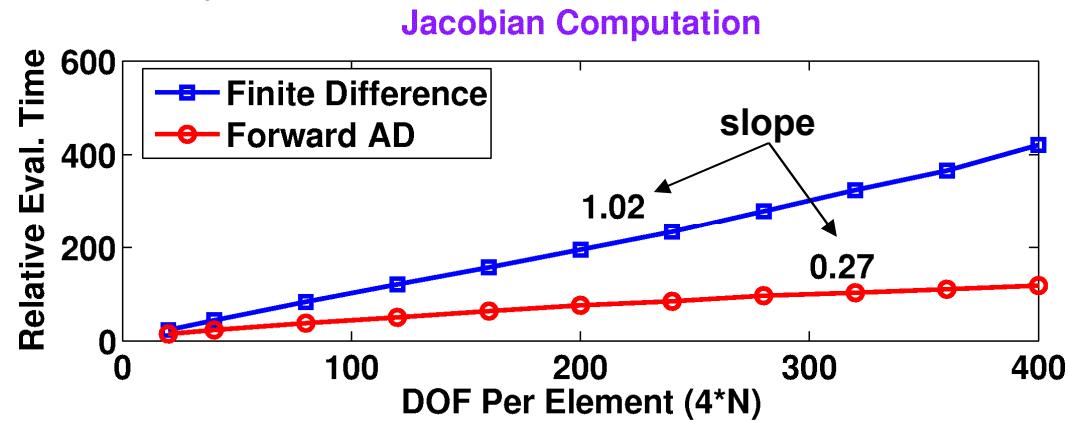
- Faster than FD
- Better scalability in number of PDEs
- Analytic derivative
- Provides Jacobian for all Charon physics

- **Reverse mode AD**

- Scalable adjoint/gradient

$$J^T w = \nabla(w^T f(x))$$

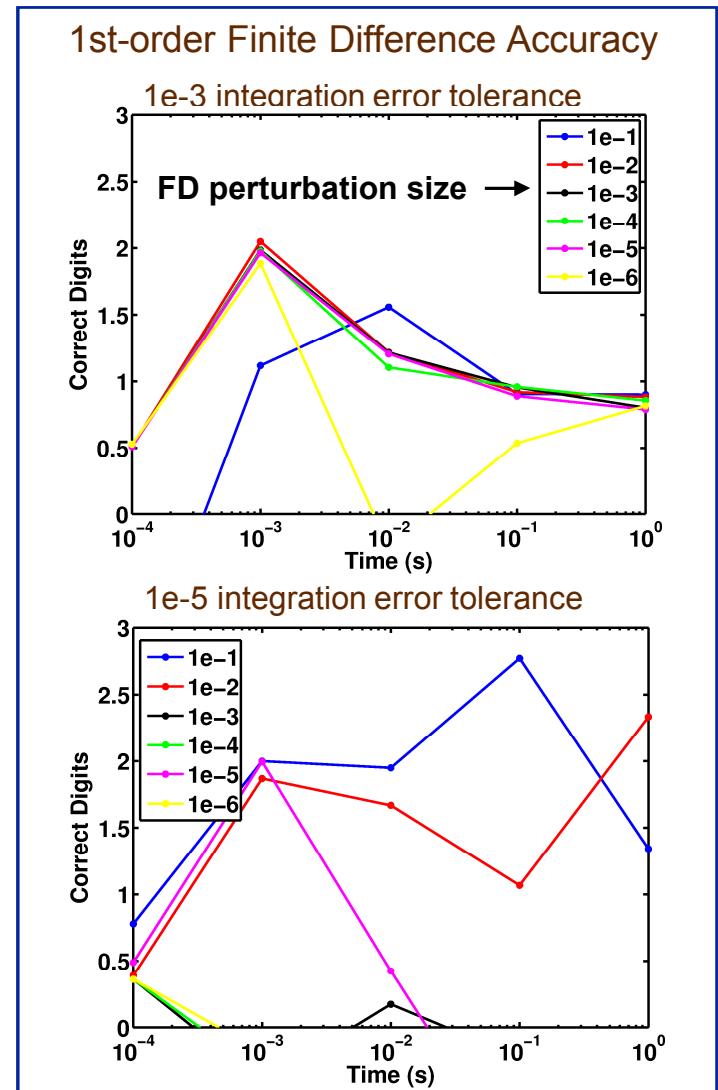
Efficiency of the element-level derivative computation



Results published, presented at ICCS 2006, Reading, UK

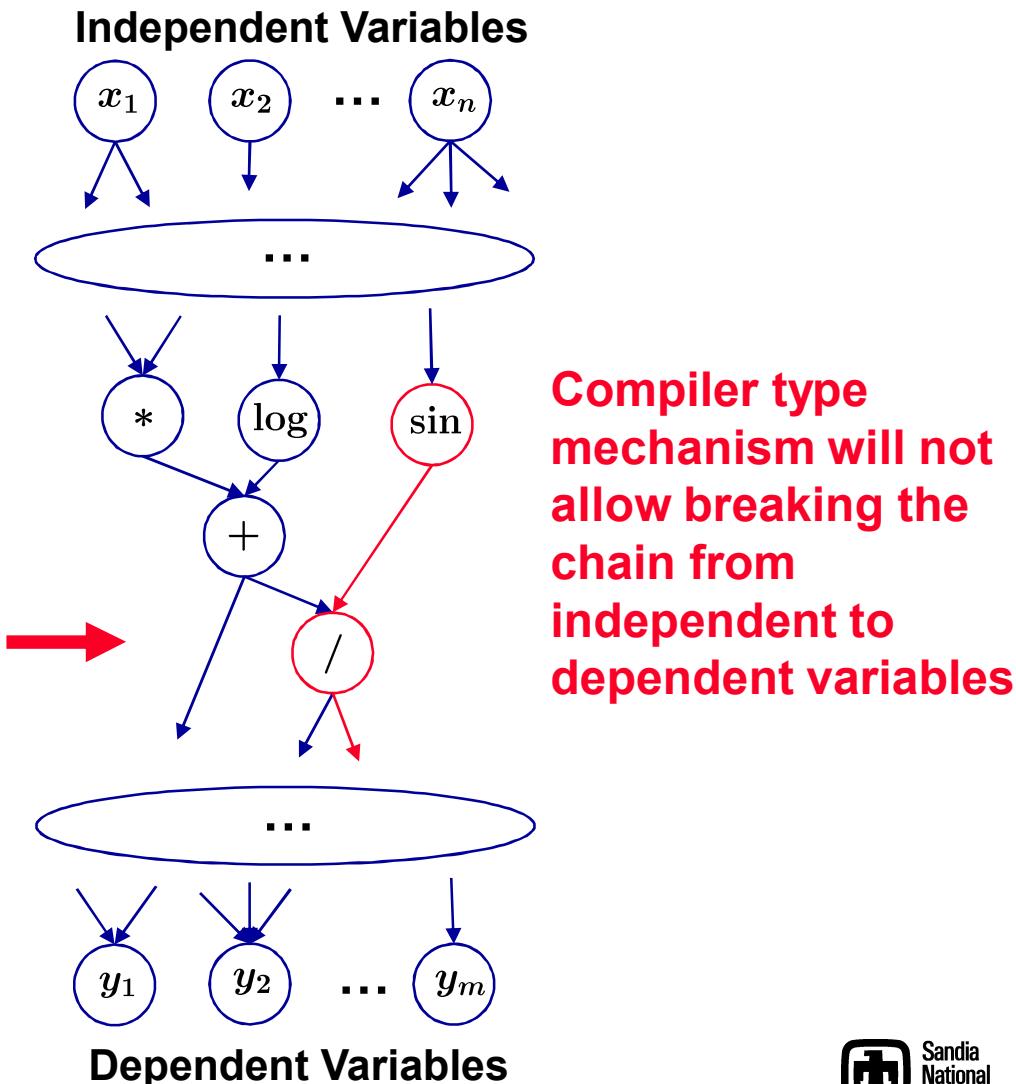
Tight Coupling Provides More Accurate, Efficient, and Robust Sensitivity Algorithm

- Run-times (1e-3 integration error):
 - Forward simulation: 105 min.
 - Rythmos sensitivities: 931 min.
 - First-order FD: ~13,000 min.
- Rythmos approach more efficient
 - 14x speed-up
- Rythmos approach more accurate
 - FD only provides square root of time integration accuracy
- Rythmos approach more robust
 - Accuracy solely dictated by time-integration accuracy
 - Picking FD perturbation size difficult



Verification of Automatic Differentiation

- Verification of the AD tools
 - Unit-test with respect to known derivatives
 - Composite tests
 - Compare to other tools
 - Compare to hand-derived
 - Compare to finite differences
- Verification of AD in application code
 - Compiler drastically simplifies this
 - All of the standard hand-coded verification techniques
 - Compare to finite differences
 - Nonlinear convergence



Charon Drift-Diffusion Formulation with Defects

Current Conservation for e- and h+	$\frac{\partial n}{\partial t} - \nabla \cdot J_n = -R_n(\psi, n, p, Y_1, \dots, Y_N), \quad J_n = -n\mu_n \nabla \psi + D_n \nabla n$
	$\frac{\partial p}{\partial t} + \nabla \cdot J_p = -R_p(\psi, n, p, Y_1, \dots, Y_N), \quad J_p = -p\mu_p \nabla \psi - D_p \nabla p$
Defect Continuity	$\frac{\partial Y_i}{\partial t} + \nabla \cdot J_{Y_i} = -R_{Y_i}(\psi, n, p, Y_1, \dots, Y_N), \quad J_{Y_i} = -\mu_i Y_i \nabla \psi - D_i \nabla Y_i$
Electric potential	$-\nabla(\varepsilon \nabla \psi(x)) = -q(p(x) - n(x) + N_D^+(x) - N_A^-(x)) - \sum_{i=1}^N q_i Y_i(x)$
Recombination/generation source terms	R_X Include electron capture and hole capture by defect species and reactions between various defect species
Electron emission/capture $Z^i \leftrightarrow Z^{i+1} + e^-$	$R_{[Z^i \rightarrow Z^{i+1} + e^-]} \propto \sigma_{[Z^i \rightarrow Z^{i+1} + e^-]} Z^i \exp\left(\frac{\Delta E_{[Z^i \rightarrow Z^{i+1} + e^-]}}{kT}\right)$ Activation Energy Cross section