

Enabling Predictive Simulation Through Embedded Sensitivity Analysis and Uncertainty Quantification

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CIS External Panel Review

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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

ELEMENT \ MATURITY	MATURITY			
	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 Minimal 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 SQE practices corrupting the simulation results?	<ul style="list-style-type: none"> Judgment only Minimal testing of any software elements Little or no SQE 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 known 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 SAs 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:

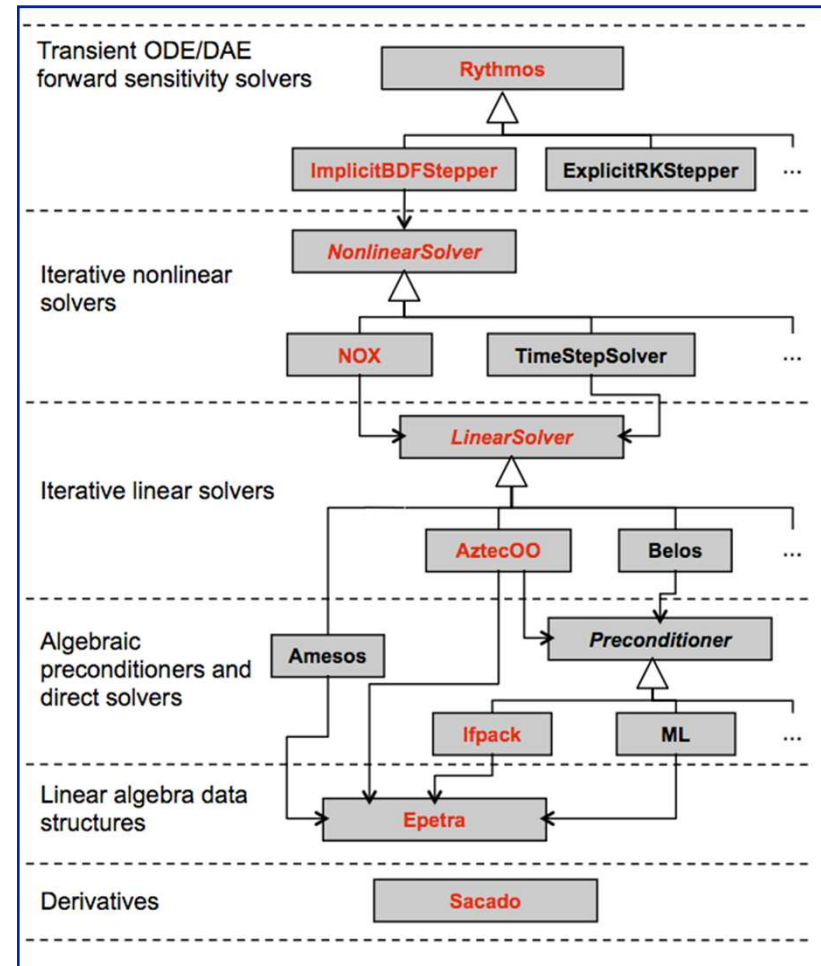
$$\frac{\partial f}{\partial \dot{x}} \left(\frac{\partial \dot{x}}{\partial p} \right) + \frac{\partial f}{\partial x} \left(\frac{\partial x}{\partial p} \right) + \frac{\partial f}{\partial p} = 0$$

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

- Achieving transformation

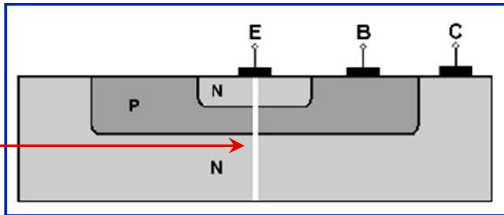
- Algorithm: Forward sensitivities
- Tight Coupling: Analytic derivatives
- Vehicle: AD + Templating

Vertical integration of Trilinos capabilities

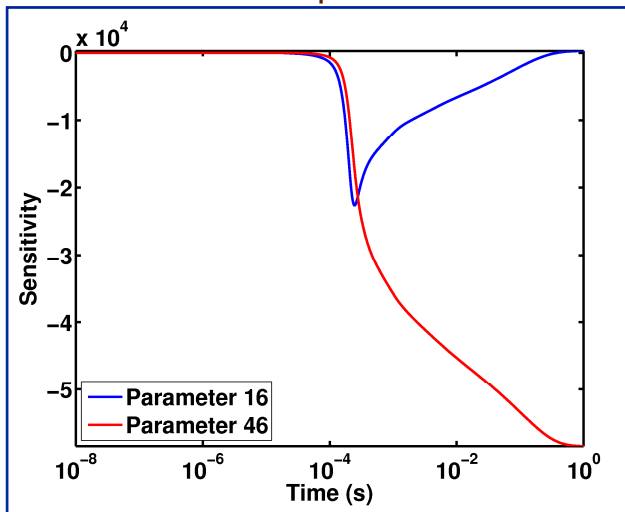


Capability Demonstrated on the QASPR Simple Prototype*

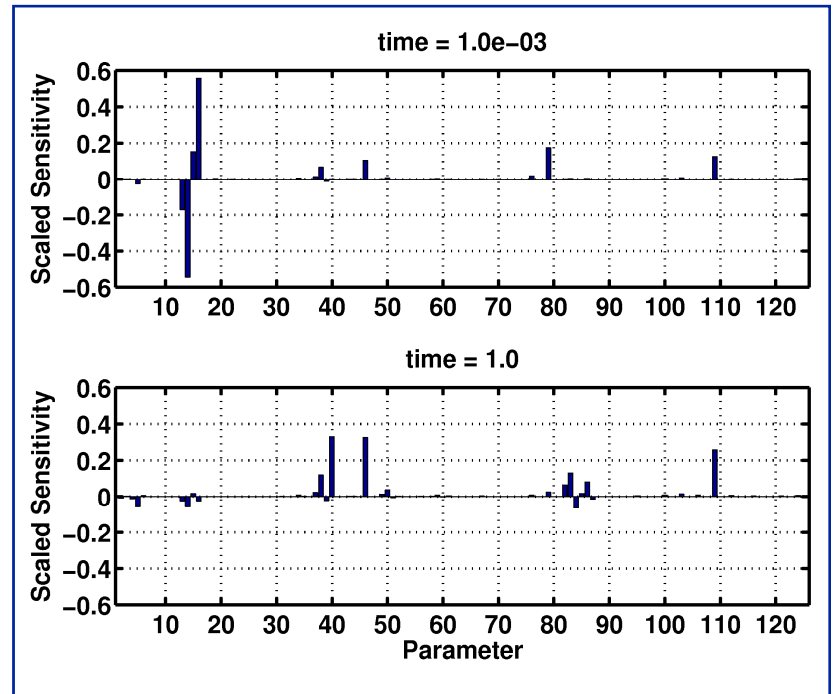
- Pseudo 1D Bipolar Junction Transistor (9x0.1 micron)
- Full defect physics
- 126 parameters



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)

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)


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MATURITY	Maturity Level 0	Maturity Level 1	Maturity Level 2	Maturity Level 3
ELEMENT				
Representation and Quantification	1. Basic representation of the problem (e.g., geometry, material properties, boundary conditions)	2. Basic quantification of the problem (e.g., error estimates, uncertainty quantification)	3. Advanced representation of the problem (e.g., multi-scale modeling, multi-physics coupling)	4. Advanced quantification of the problem (e.g., high-order accuracy, adaptive refinement)
Physics and Material Model Fidelity	1. Basic physics and material model fidelity (e.g., linear elasticity, isotropic material properties)	2. Basic physics and material model fidelity (e.g., nonlinear elasticity, anisotropic material properties)	3. Advanced physics and material model fidelity (e.g., multi-scale modeling, multi-physics coupling)	4. Advanced physics and material model fidelity (e.g., high-order accuracy, adaptive refinement)
Code Verification	1. Basic code verification (e.g., unit tests, integration tests)	2. Basic code verification (e.g., unit tests, integration tests)	3. Advanced code verification (e.g., multi-scale modeling, multi-physics coupling)	4. Advanced code verification (e.g., high-order accuracy, adaptive refinement)
Solution Verification	1. Basic solution verification (e.g., convergence studies, error estimates)	2. Basic solution verification (e.g., convergence studies, error estimates)	3. Advanced solution verification (e.g., multi-scale modeling, multi-physics coupling)	4. Advanced solution verification (e.g., high-order accuracy, adaptive refinement)
Model Validation	1. Basic model validation (e.g., comparison with experimental data)	2. Basic model validation (e.g., comparison with experimental data)	3. Advanced model validation (e.g., multi-scale modeling, multi-physics coupling)	4. Advanced model validation (e.g., high-order accuracy, adaptive refinement)
Uncertainty Quantification and Sensitivity Analysis	1. Basic uncertainty quantification and sensitivity analysis (e.g., Monte Carlo simulation)	2. Basic uncertainty quantification and sensitivity analysis (e.g., Monte Carlo simulation)	3. Advanced uncertainty quantification and sensitivity analysis (e.g., multi-scale modeling, multi-physics coupling)	4. Advanced uncertainty quantification and sensitivity analysis (e.g., high-order accuracy, adaptive refinement)

Impacting CIS Transformation Mission

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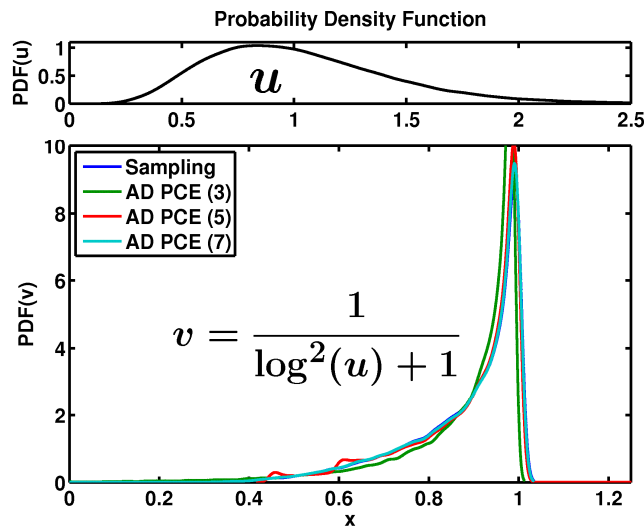
Templating Enables Automatic Embedded Uncertainty Propagation Research

Stochastic Galerkin method

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

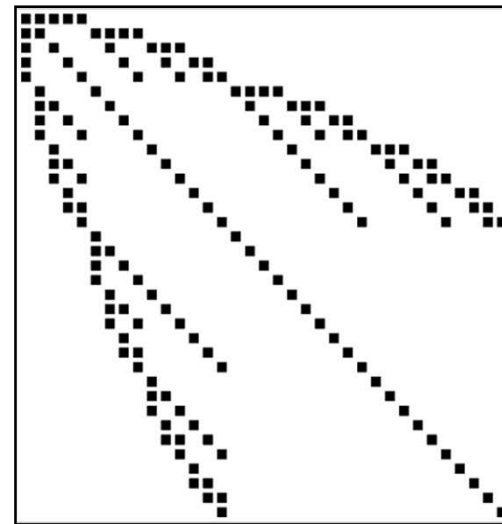
Embedded SG techniques using AD-like procedure (completed)

– **Compute projection op-by-op**



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

– **Linear algebra research**



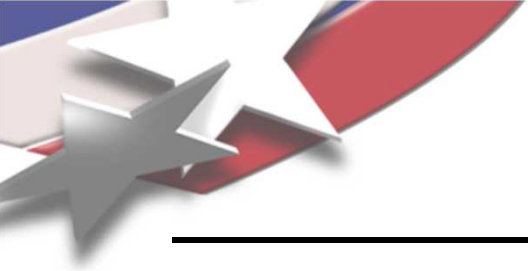
Encapsulated in new Trilinos package *Stokhos*

– **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

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

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

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

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

- **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:**

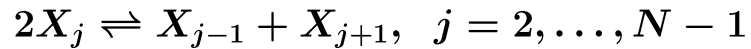
$$x(t) = \sum_{k=0}^d x_k t^k \longrightarrow \sum_{k=0}^d f_k t^k = f(x(t)) + O(t^{d+1}), \quad f_k = \frac{1}{k!} \frac{d^k}{dt^k} f(x(t))$$

- **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: Efficiency of the element-level derivative computation



Steady-state mass transfer equations:

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

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

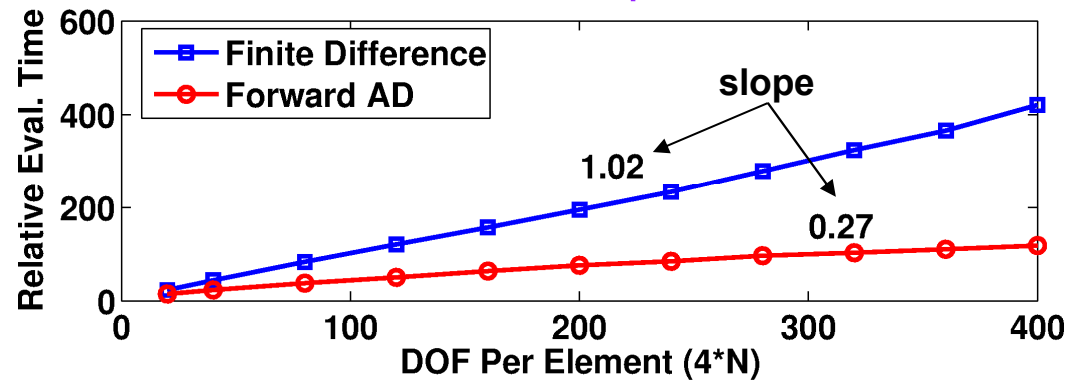
- **Forward mode AD**

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

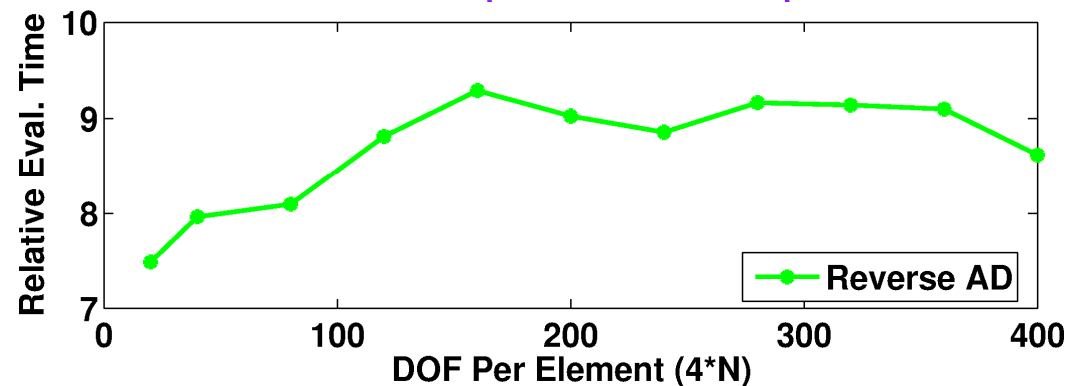
- **Reverse mode AD**

- Scalable adjoint/gradient
- $$J^T w = \nabla(w^T f(x))$$

Jacobian Computation



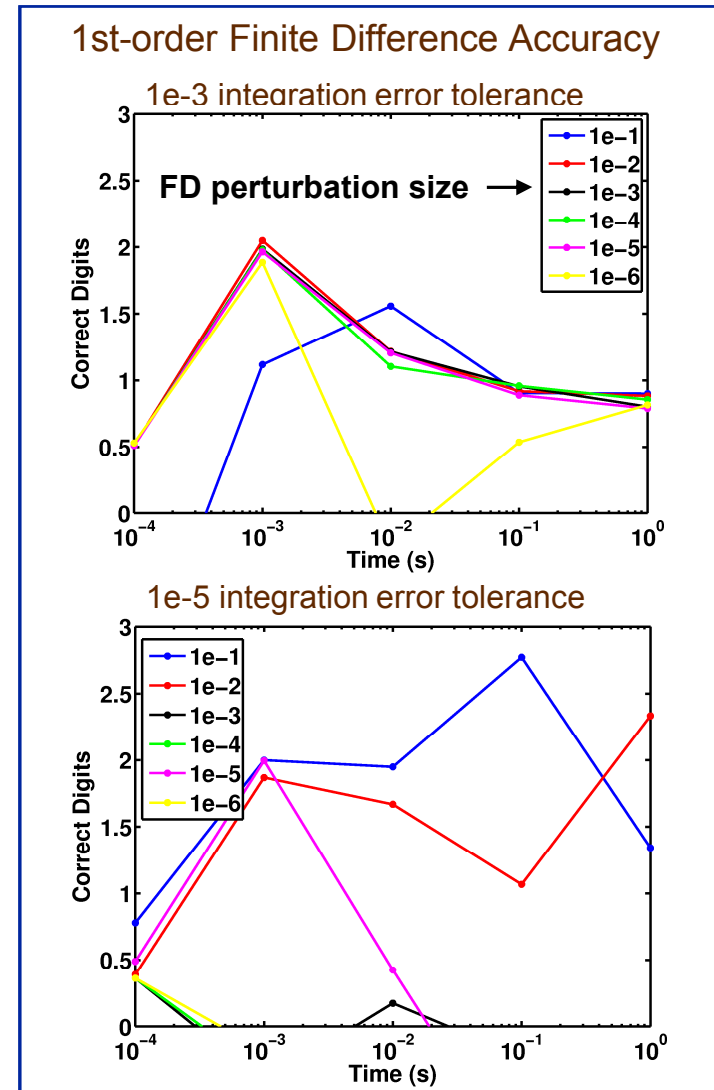
Jacobian-Transpose Product Computation



Results published, presented at ICCS 2006, Reading, UK

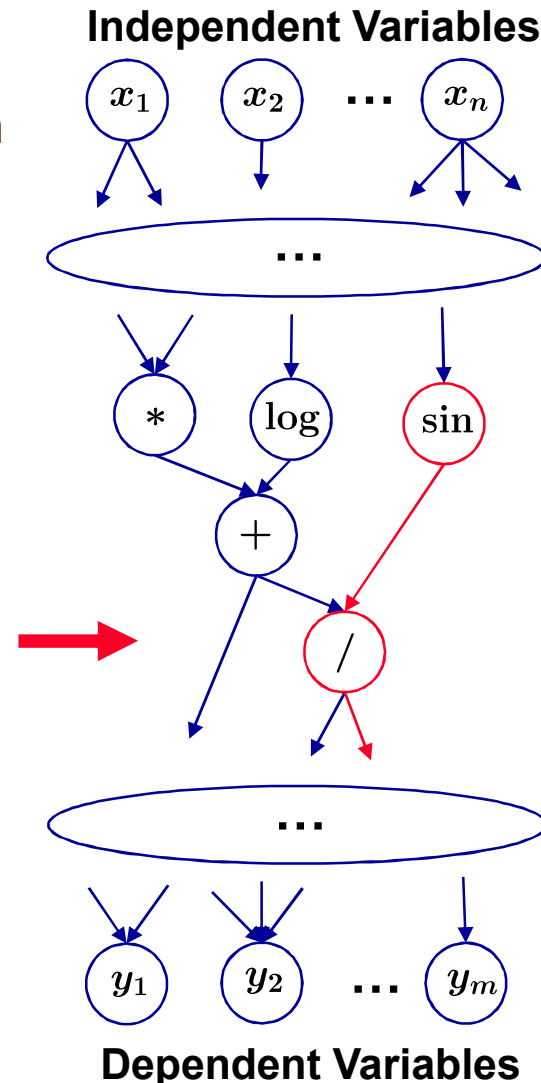
Tight Coupling Provides More Accurate, Efficient, and Robust Sensitivity Algorithm

- 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



Compiler type mechanism will not allow breaking the chain from independent to dependent variables

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