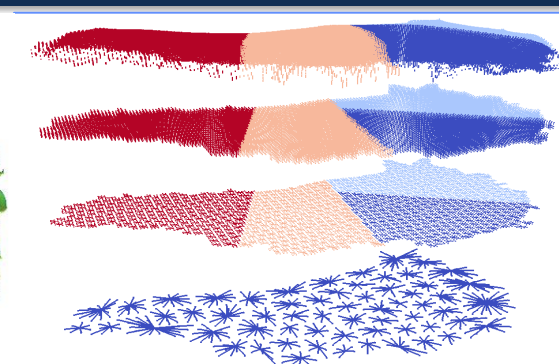
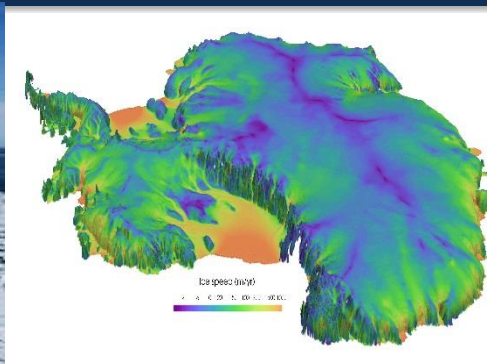


Exceptional service in the national interest



The *Albany/FELIX* Land-Ice Dynamical Core

I. Tezaur¹, A. Salinger¹, M. Perego¹, R. Tuminaro¹, J. Jakeman¹, M. Eldred¹, J. Watkins¹,
S. Price², I. Demeshko²

¹ Sandia National Laboratories, Albuquerque, NM and Livermore, CA, USA.

² Los Alamos National Laboratory, Los Alamos, NM, USA.

AUG 2017 Livermore, CA Jan. 17-18, 2017



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The PISCEES Project

* Proposal for *follow-up funding* under SciDaC4 in preparation.



PISCEES
SciDaC3 Application Partnership
 (DOE's BER + ASCR divisions)
 June 2012 – 2017*

FSU FELIX
 FSU
 Finite Element
 Full Stokes Model

Albany/FELIX
 SNL
 Finite Element
 First-Order Stokes Model

BISICLES
 LBNL
 Finite Volume
 L1L2 Model

3 land-ice
 dycores
 developed
 under
 PISCEES

↑
 Increased
 fidelity



PISCEES: Predicting Ice Sheet Climate Evolution at Extreme Scales
FELIX: Finite Elements for Land Ice eXperiments
BISICLES: Berkeley Ice Sheet Initiative for Climate at Extreme Scales



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Goal: support DOE
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Sandia's Role in PISCEES

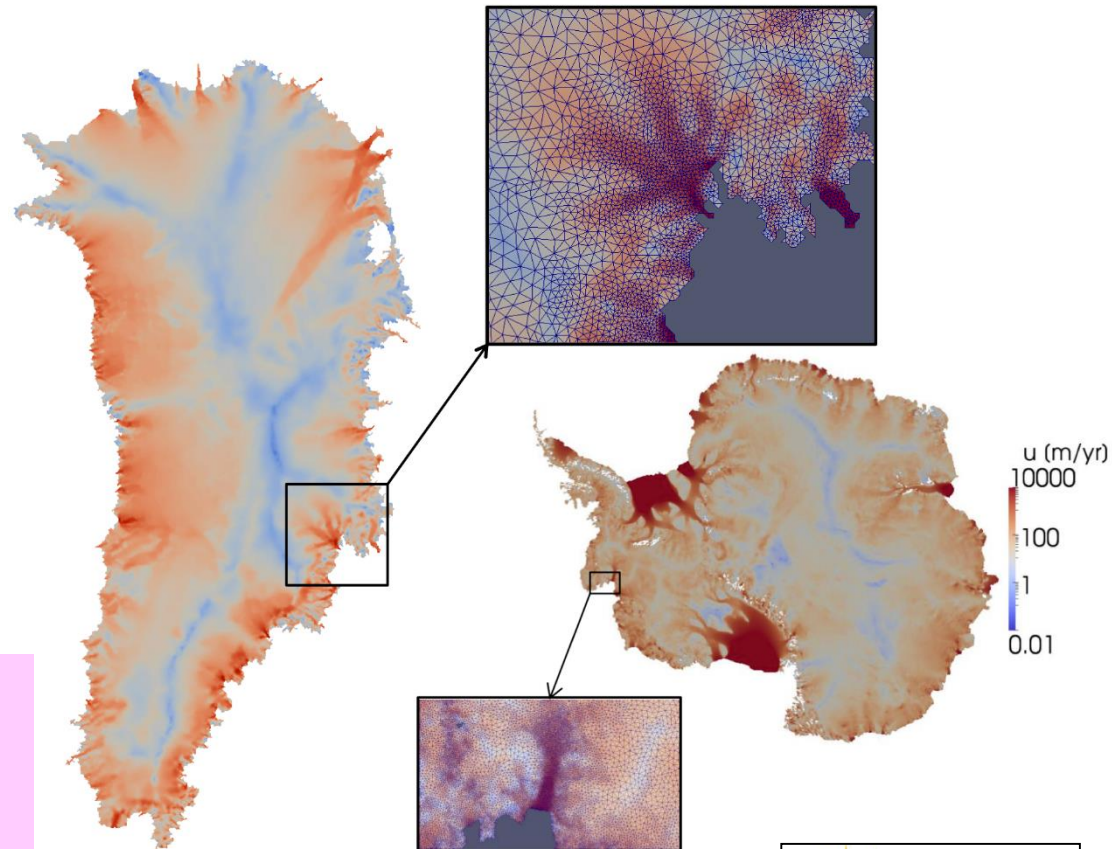
Sandia's Role in the PISCEES Project: to **develop** and **support** a robust and scalable land ice solver based on the "First-Order" (FO) Stokes equations → *Albany/FELIX**

Requirements for *Albany/FELIX*:

- **Unstructured grid** finite elements.
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- **Verified** and **validated**.
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- **Advanced analysis** capabilities: deterministic inversion, calibration, uncertainty quantification.

As part of **ACME DOE earth system model**, solver will provide actionable predictions of 21st century sea-level rise (including uncertainty).

*Finite Elements for Land Ice eXperiments



 Albany



Sandia's Role in PISCEES

*Albany/FELIX =
production code!*



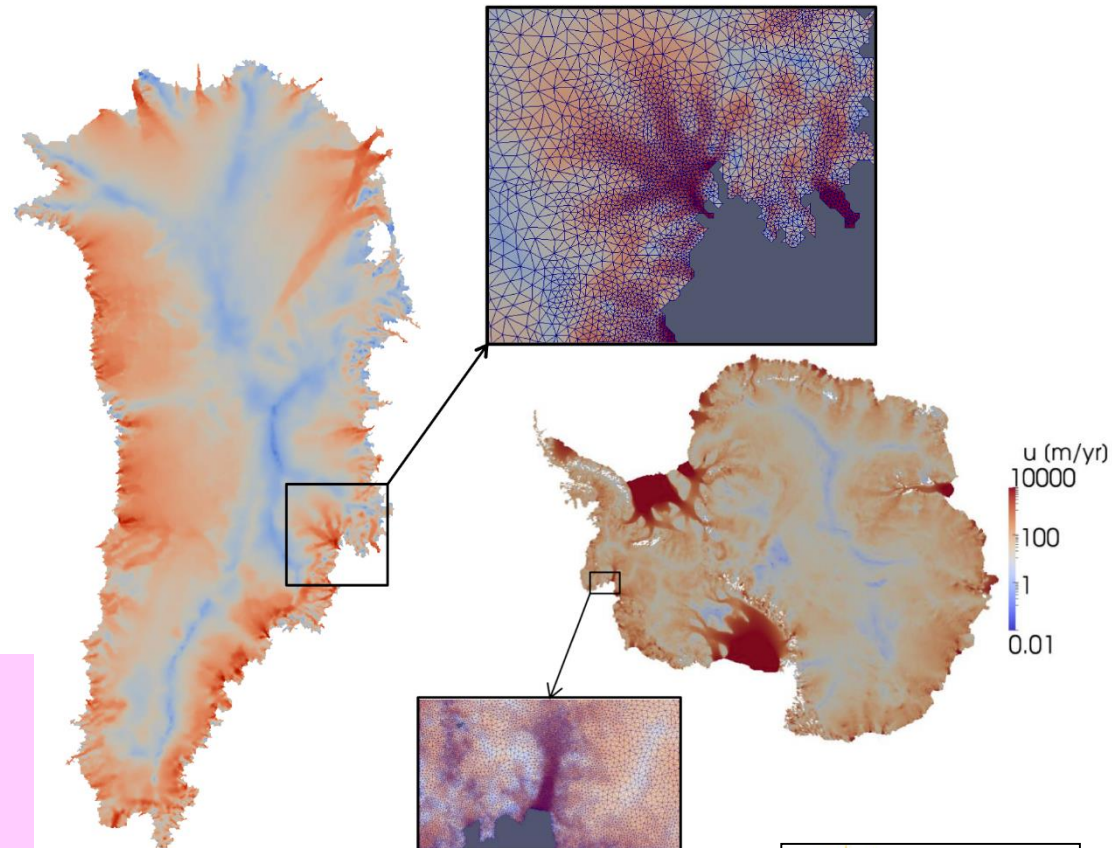
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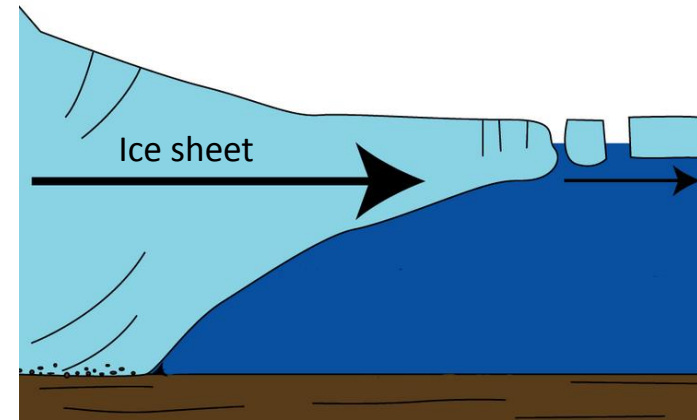
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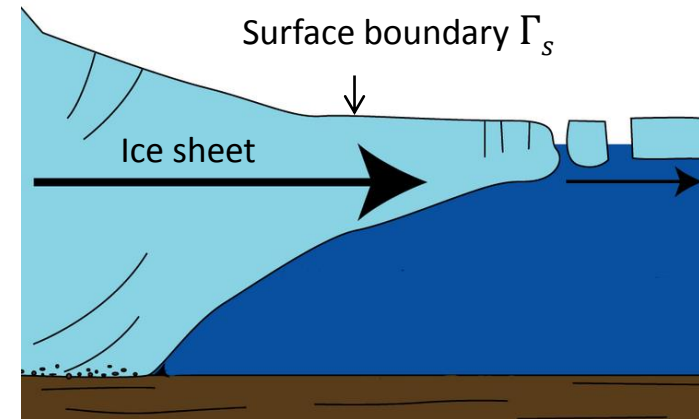
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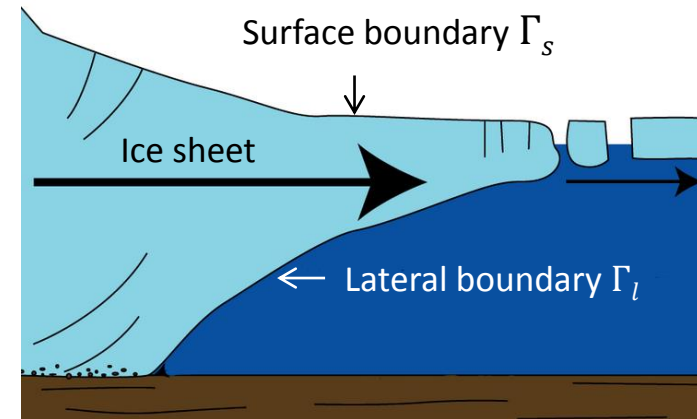
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$$2\mu \dot{\epsilon}_i \cdot \mathbf{n} = \begin{cases} \rho g z \mathbf{n}, & \text{if } z > 0 \\ 0, & \text{if } z \leq 0 \end{cases}, \quad \text{on } \Gamma_l$$



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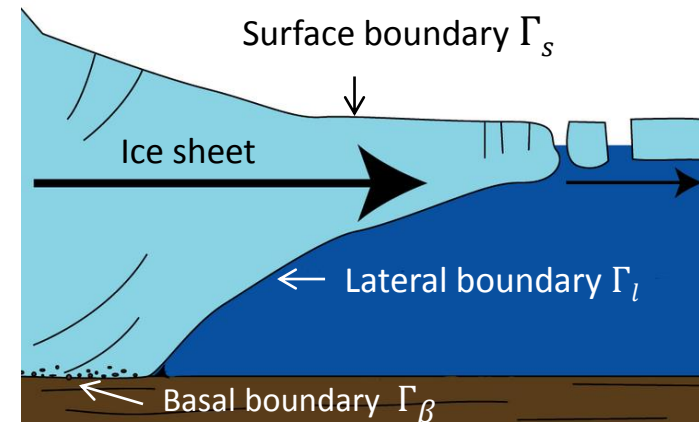
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- **Basal sliding BC:** $2\mu \dot{\epsilon}_i \cdot \mathbf{n} + \beta(x, y) u_i = 0$, on Γ_β



$$\beta(x, y) = \text{basal sliding coefficient}$$

Thickness & Temperature Equations

- Model for **evolution of the boundaries** (thickness evolution equation):

$$\frac{\partial H}{\partial t} = -\nabla \cdot (\bar{\mathbf{u}}H) + \dot{b}$$

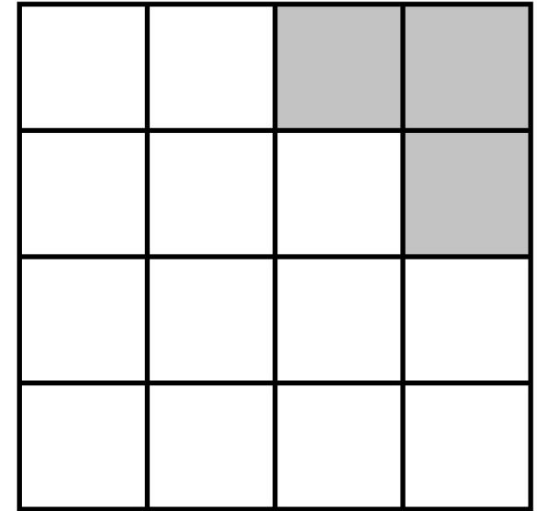
where $\bar{\mathbf{u}}$ = vertically averaged velocity, \dot{b} = surface mass balance (conservation of mass).

- **Temperature equation** (advection-diffusion):

$$\rho c \frac{\partial T}{\partial t} = \nabla \cdot (k\nabla T) - \rho c \mathbf{u} \cdot \nabla T + 2\dot{\epsilon}\sigma$$

(energy balance).

- **Flow factor** A in Glen's law depends on temperature T :
 $A = A(T)$.
- Ice sheet **grows/retreats** depending on thickness H .



time t_0

Ice-covered ("active")
cells shaded in white
($H > H_{min}$)

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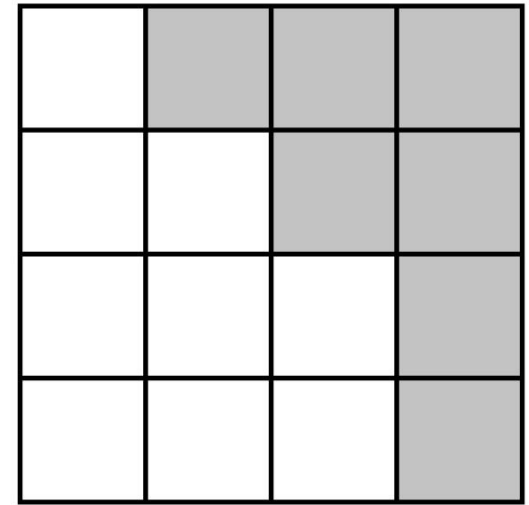
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time t_1

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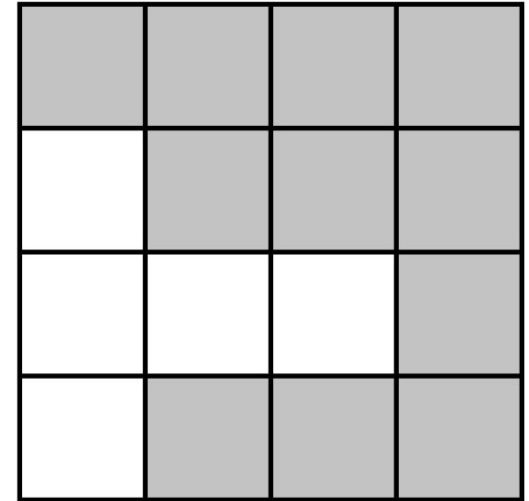
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time t_2

Ice-covered ("active")
cells shaded in white
($H > H_{min}$)

Summary of Ice Sheet Equations & Codes Sandia National Laboratories

Momentum Balance: First-Order Stokes PDEs

$$\begin{cases} -\nabla \cdot (2\mu \dot{\epsilon}_1) = -\rho g \frac{\partial s}{\partial x} \\ -\nabla \cdot (2\mu \dot{\epsilon}_2) = -\rho g \frac{\partial s}{\partial y} \end{cases}, \quad \text{in } \Omega$$

with **Glen's law** viscosity $\mu = \frac{1}{2} A(T)^{-\frac{1}{3}} \left(\frac{1}{2} \sum_{ij} \dot{\epsilon}_{ij}^2 \right)^{\left(\frac{2}{3}\right)}$.

Conservation of Mass: thickness evolution PDE

$$\frac{\partial h}{\partial t} = -\nabla \cdot (\bar{u}h) + \dot{b}$$

Energy Balance: temperature advection-diffusion PDE

$$\rho c \frac{\partial T}{\partial t} = \nabla \cdot (k \nabla T) - \rho c \mathbf{u} \cdot \nabla T + 2 \dot{\epsilon} \sigma$$

Code:

C++

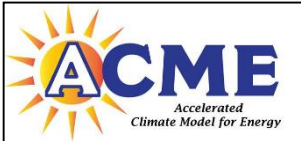


Albany/FELIX



CISM/Albany
MPAS/Albany

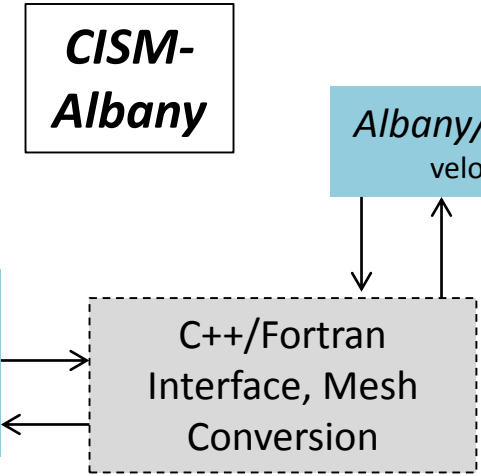
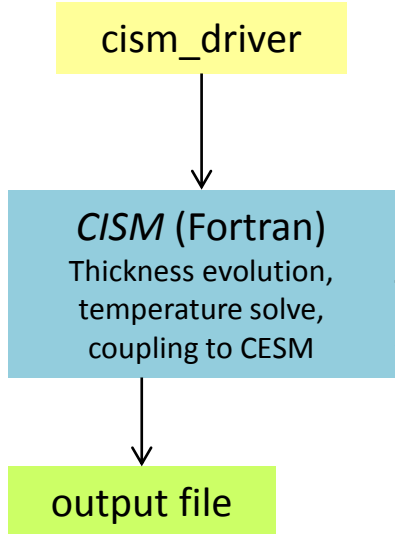
Fortran



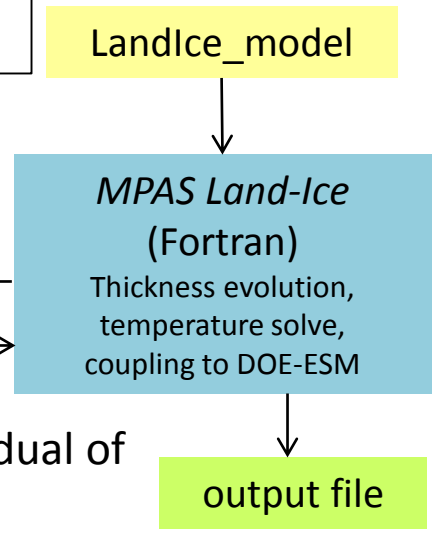
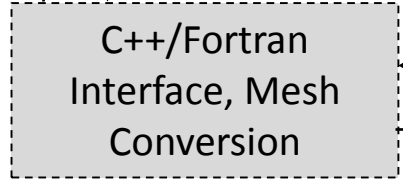
CISM-Albany and MPAS-Albany

CISM-Albany

MPAS-Albany

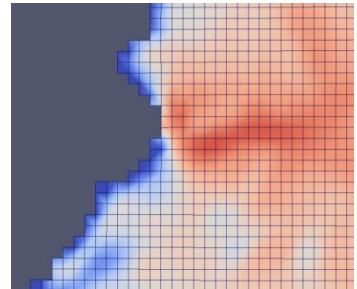


Albany/FELIX (C++)
velocity solve

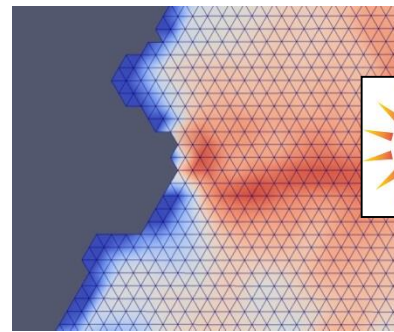
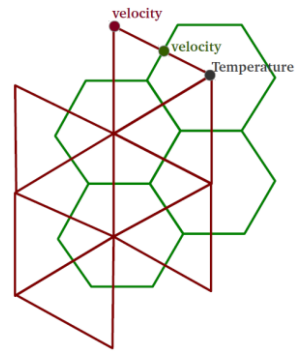


CESM
COMMUNITY EARTH SYSTEM MODEL

- Structured hexahedral meshes (rectangles extruded to hexes).



- Tetrahedral meshes (dual of hexagonal mesh, extruded to tets).



ACME
Accelerated
Climate Model for Energy

Albany/FELIX has been coupled to two land ice dycores: **Community Ice Sheet Model (CISM)** and **Model for Prediction Across Scales for Land-Ice (MPAS)**

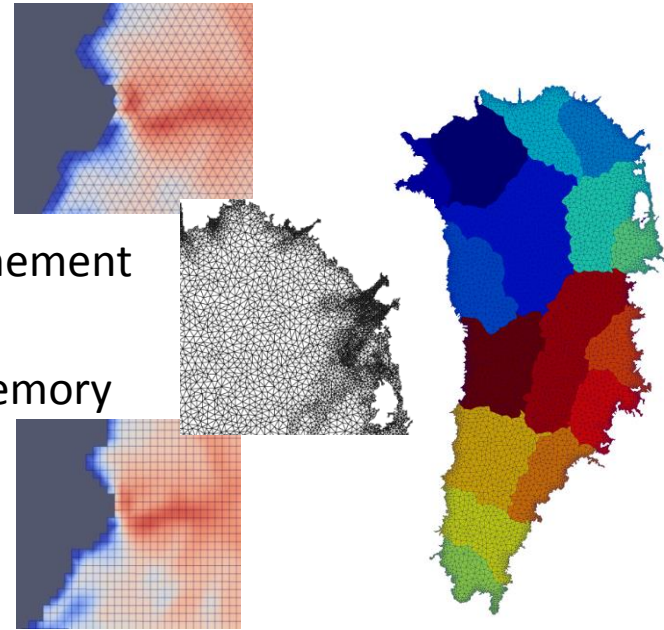
Meshes and Data



Meshes and Data

Meshes: can use any mesh but interested specifically in

- **Structured hexahedral** meshes (compatible with *CISM*).
- **Uniform tetrahedral** meshes (compatible with *MPAS*)
- **Unstructured Delaunay triangle** meshes w/ regional refinement based on gradient of surface velocity.
- **Unstructured adaptively-refined** meshes generated in memory using **PAALS*** (Parallel Albany Adaptive Loop w/ SCOREC).

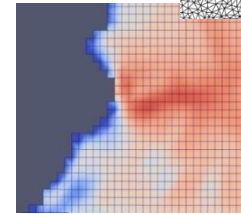
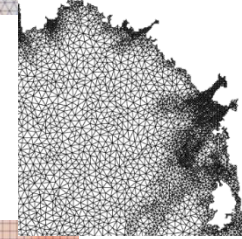
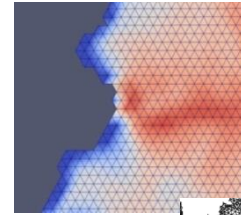


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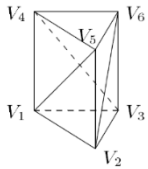
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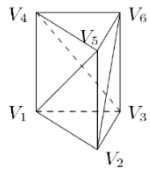
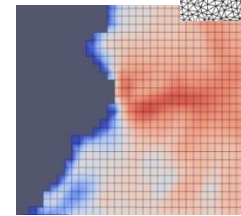
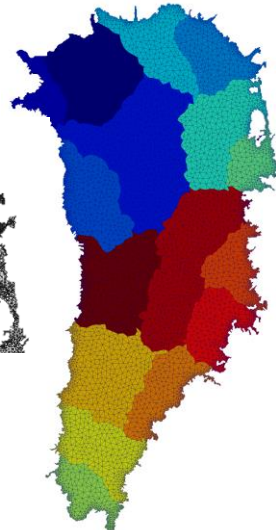
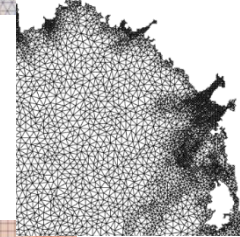
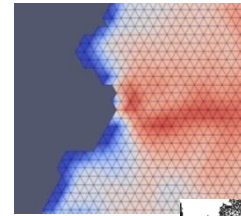
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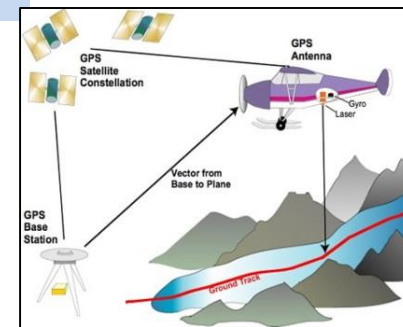
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Data: needs to be imported into code to run “real” problems (Greenland, Antarctica).

- **Surface data** are available from measurements (satellite infrarometry, radar, altimetry): ice extent, surface topography, surface velocity, surface mass balance.
- **Interior ice data** (ice thickness, basal friction) cannot be measured; estimated by solving an inverse problem.

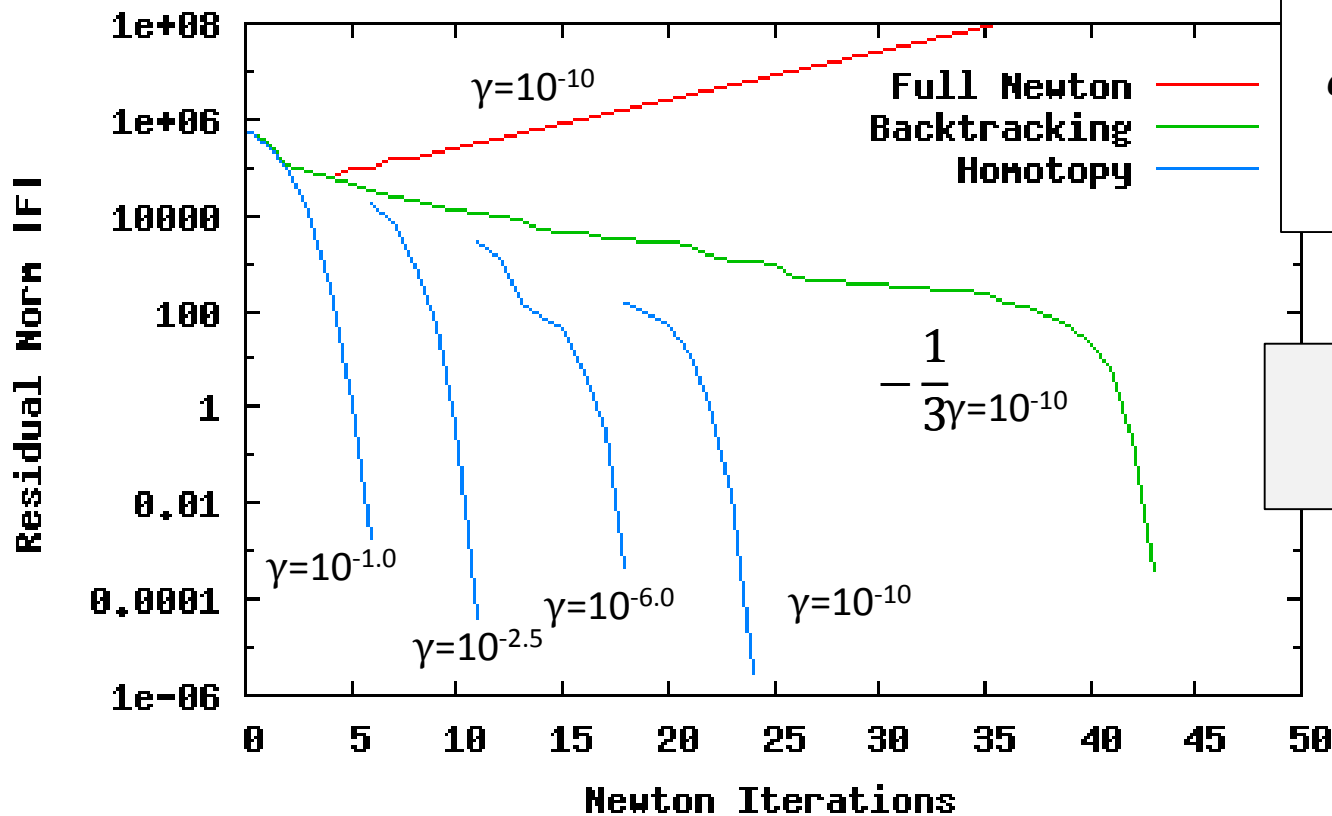


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Performance: Robustness and Scalability



Robustness of Newton's Method via Homotopy Continuation (LOCA)



$$\dot{\epsilon}_1^T = (2\dot{\epsilon}_{11} + \dot{\epsilon}_{22}, \dot{\epsilon}_{12}, \dot{\epsilon}_{13})$$

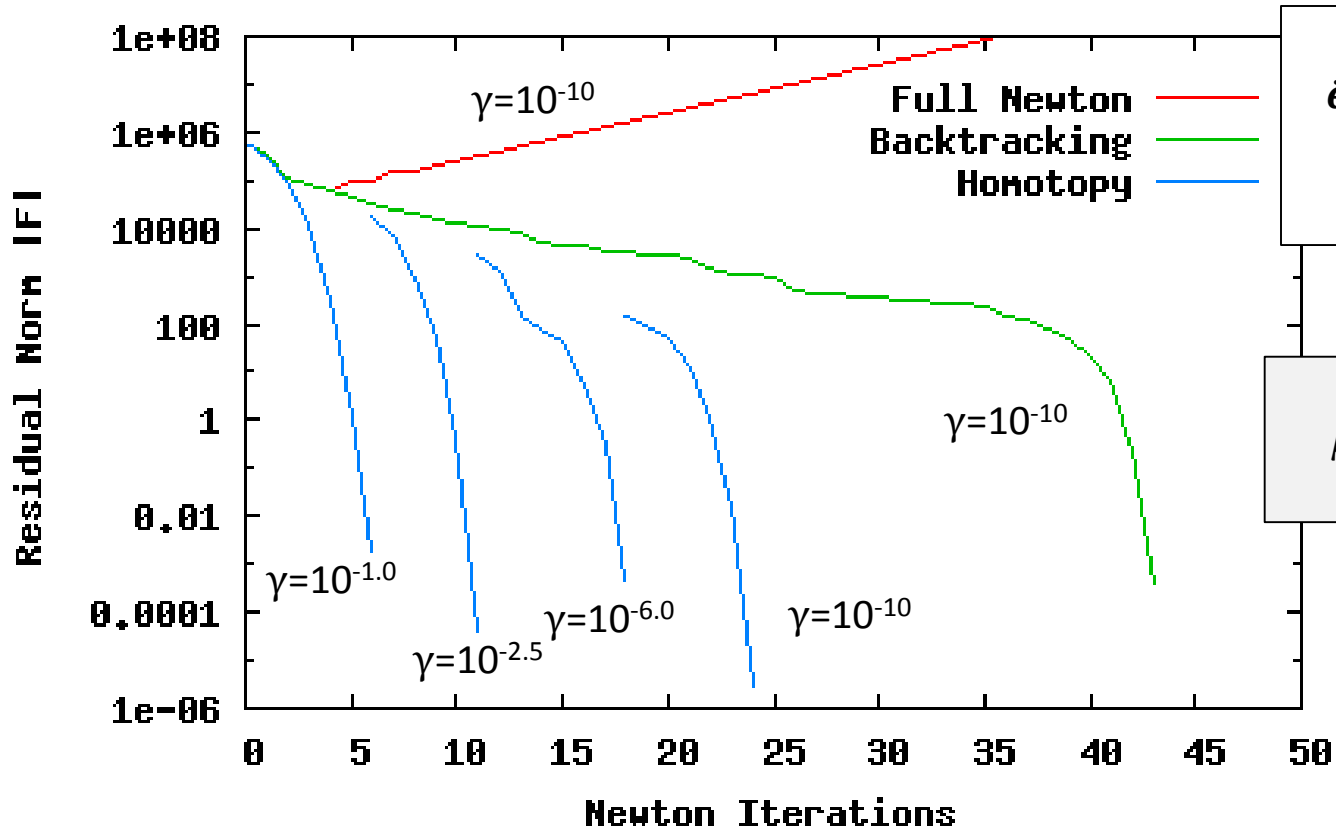
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Glen's Law Viscosity:

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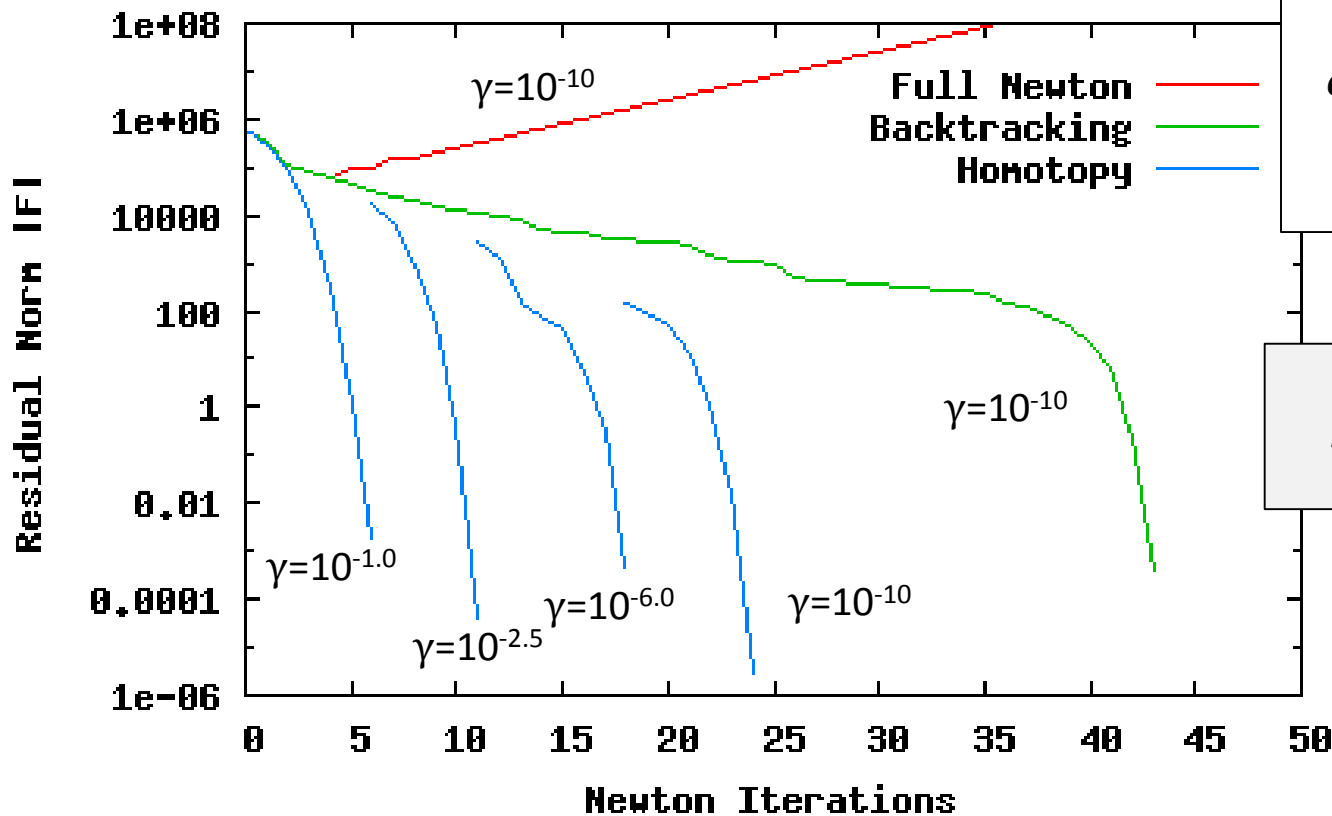
$$\dot{\epsilon}_{ij} = \frac{1}{2} \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right)$$

Glen's Law Viscosity:

$$\mu = \frac{1}{2} A^{-\frac{1}{3}} \left(\frac{1}{2} \sum_{ij} \dot{\epsilon}_{ij}^2 + \gamma \right)^{-\frac{2}{3}}$$

γ = regularization parameter

Robustness of Newton's Method via Homotopy Continuation (LOCA)



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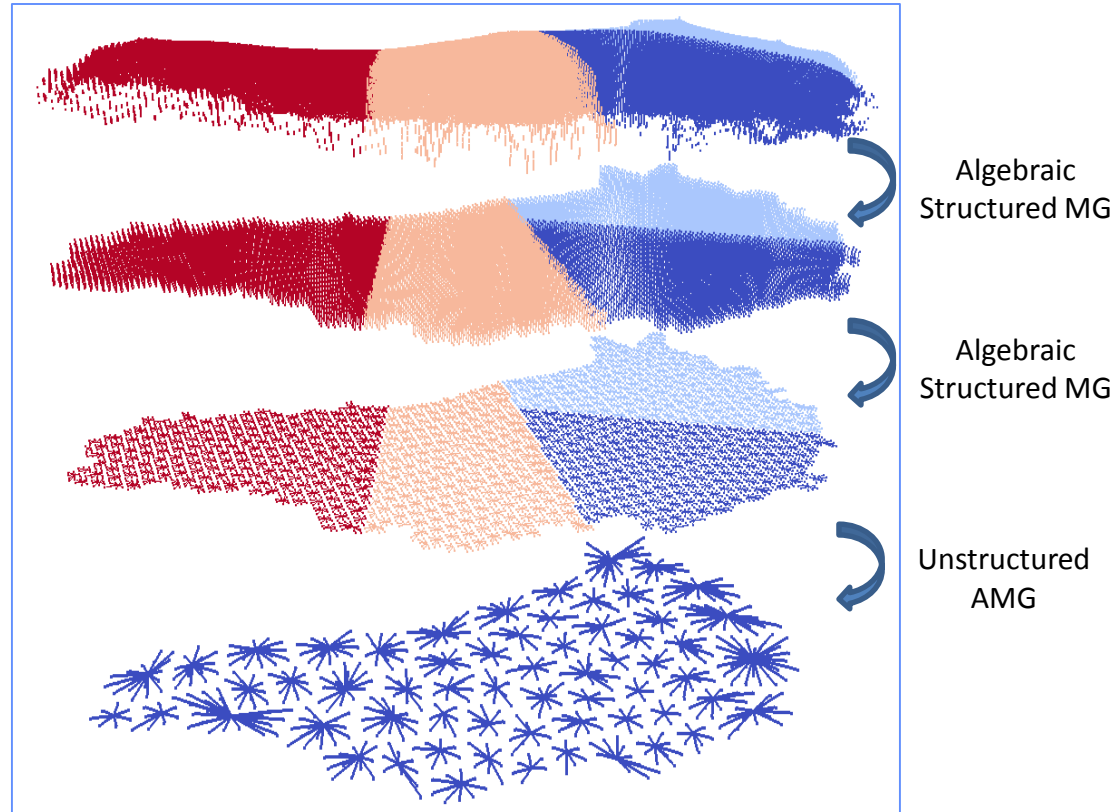
Newton most robust with full step + homotopy
continuation of $\gamma \rightarrow 10^{-10}$: **converges out-of-the-box!**

Scalability via Algebraic Multi-Grid Preconditioning with Semi-Coarsening

Bad aspect ratios ($dx \gg dz$) ruin classical AMG convergence rates!

- relatively small horizontal coupling terms, hard to smooth horizontal errors
- ⇒ Solvers (AMG and ILU) must take aspect ratios into account

New AMG solver based on aggressive **semi-coarsening** has been developed by R. Tuminaro (available in *ML/MueLu* packages of *Trilinos*)



See (Tezaur *et al.*, 2015),
(Tuminaro *et al.*, 2016).

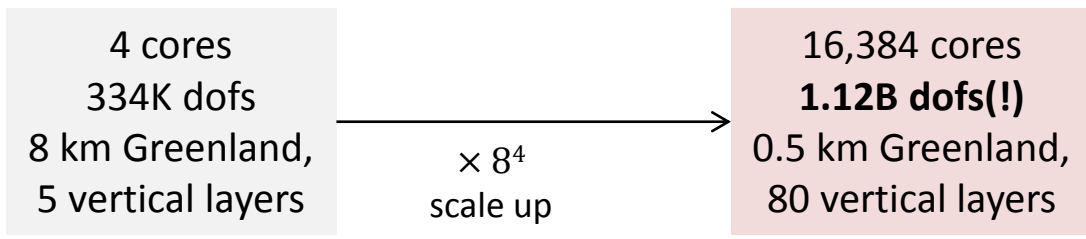
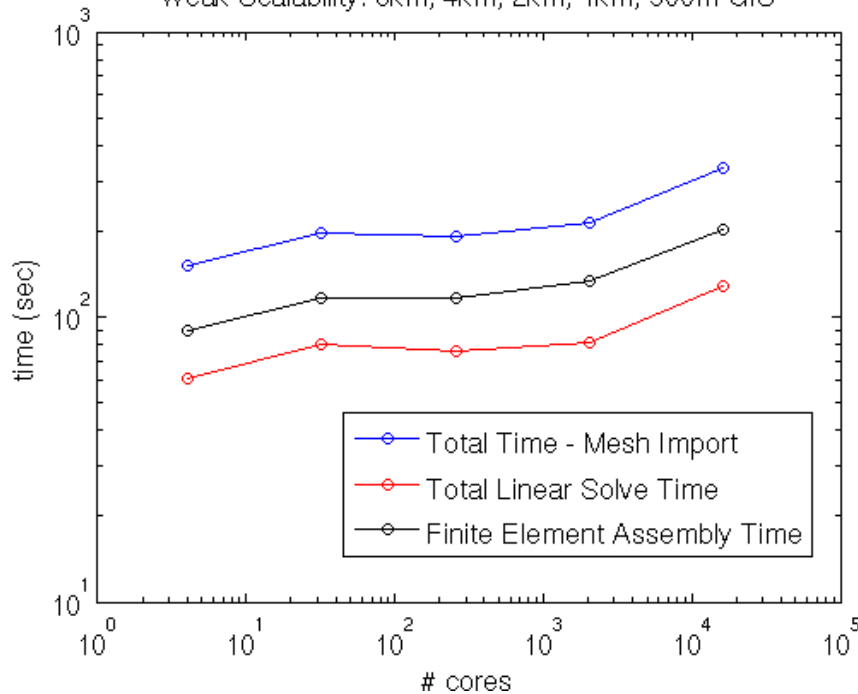
Scalability studies (next slides):
New AMG preconditioner vs. ILU



Greenland Controlled Weak Scalability Study

New AMG preconditioner

Weak Scalability: 8km, 4km, 2km, 1km, 500m GIS

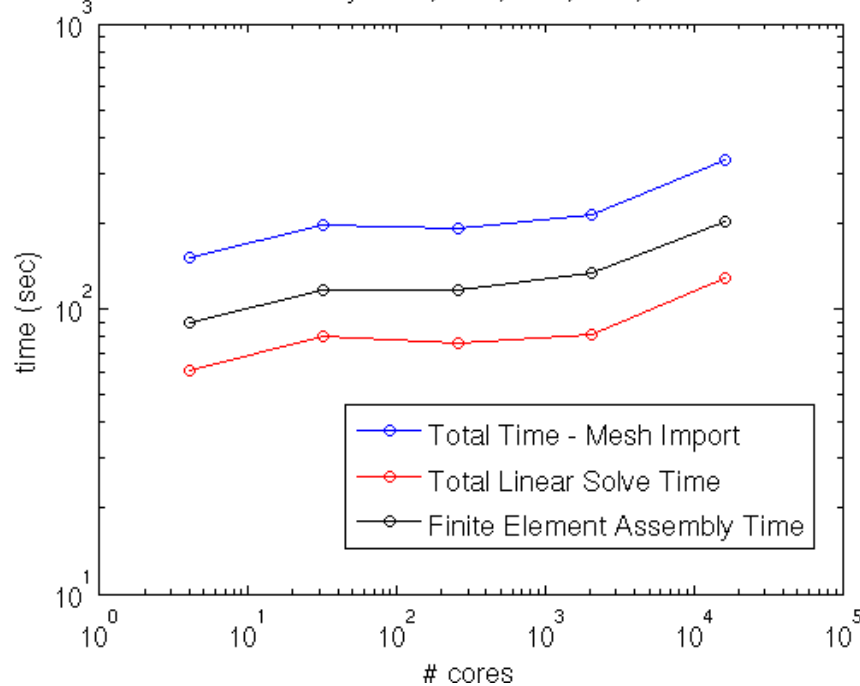


- Weak scaling study with fixed dataset, 4 mesh bisections.
- ~70-80K dofs/core.
- **Conjugate Gradient (CG) iterative method** for linear solves (faster convergence than GMRES).
- **New AMG preconditioner** developed by R. Tuminaro based on **semi-coarsening** (coarsening in z-direction only).
- **Significant improvement** in scalability with new AMG preconditioner over ILU preconditioner!

Greenland Controlled Weak Scalability Study

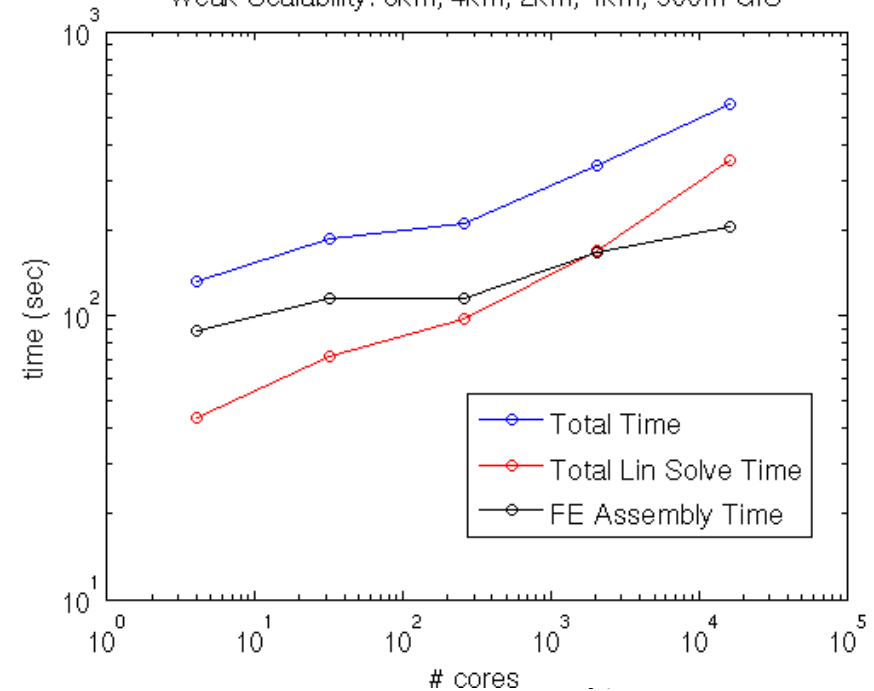
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ILU preconditioner

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4 cores
334K dofs
8 km Greenland,
5 vertical layers

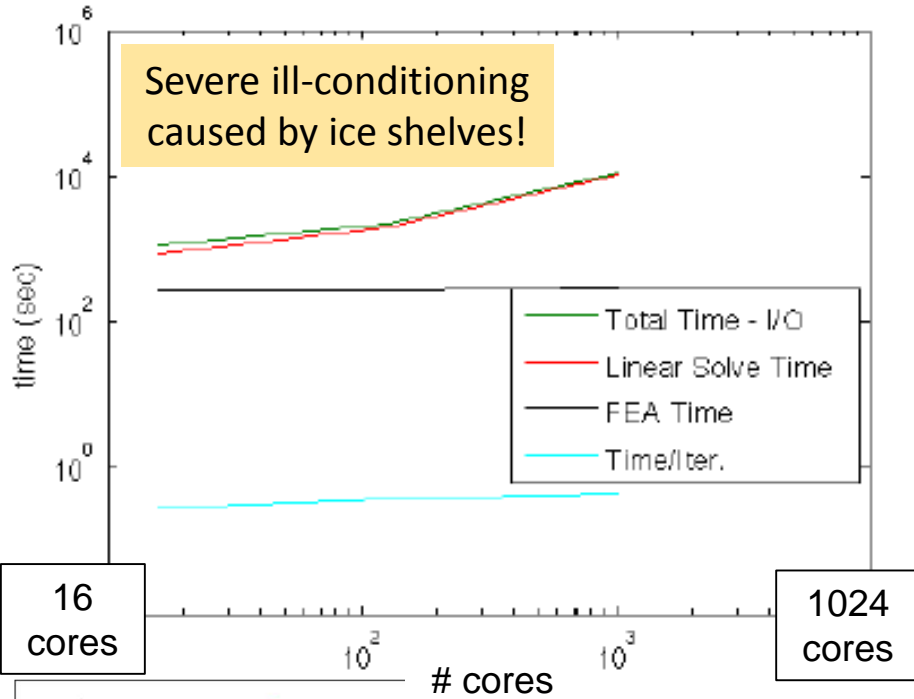
→
× 8⁴
scale up

16,384 cores
1.12B dofs(!)
0.5 km Greenland,
80 vertical layers

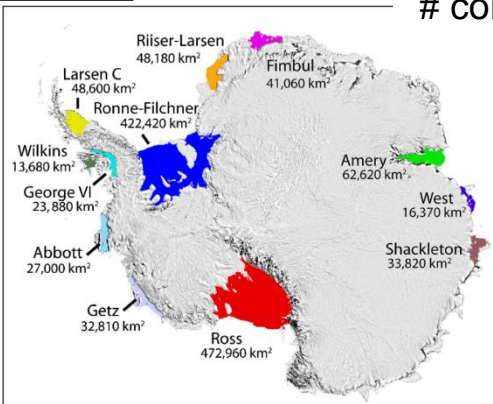
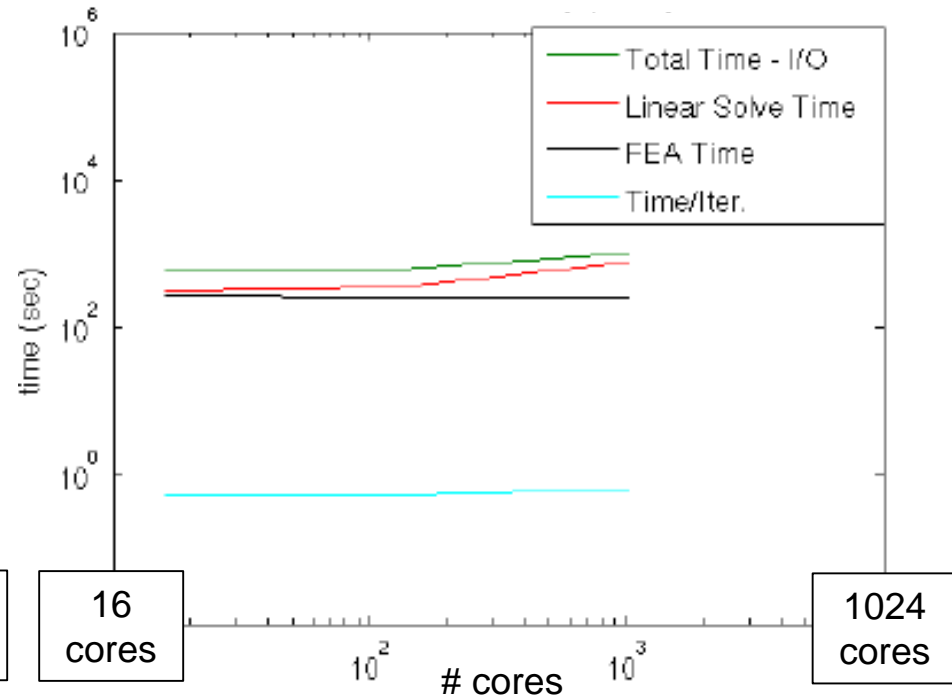
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Antarctica Weak Scalability Study

ILU preconditioner



New AMG preconditioner preconditioner



AMG preconditioner *less sensitive to ill-conditioning* caused by *ice shelves* than ILU (ice shelves → Green's function with modest horizontal decay → ILU is less effective).

Performance Portability



Performance Portability via *Kokkos*

We need to be able to run *Albany/FELIX* on ***new architecture machines*** (hybrid systems) and ***manycore devices*** (multi-core CPU, NVIDIA GPU, Intel Xeon Phi, etc.)

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→ achieved using the ***Kokkos*** library/programming model.

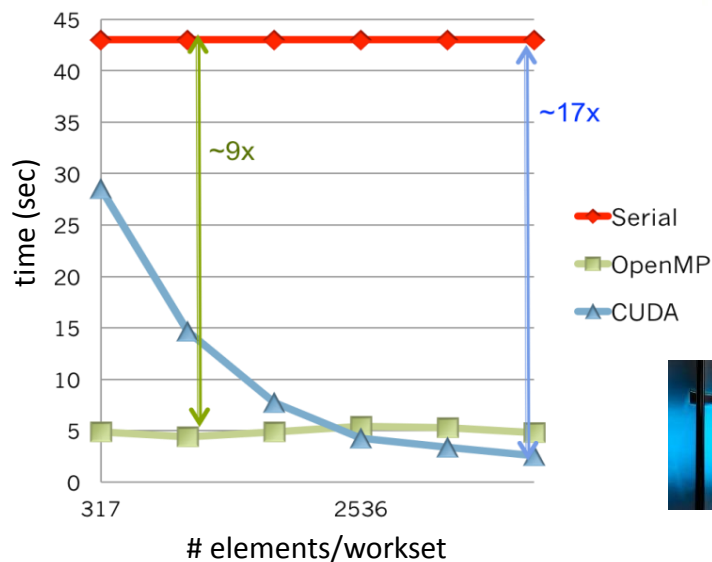
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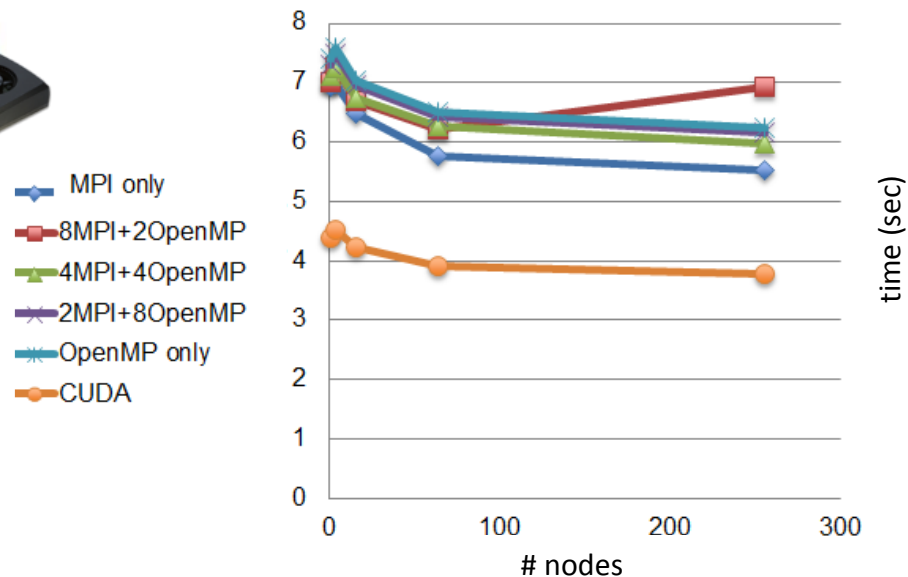
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FEA Time (Greenland, on Shannon)



Compute Time (Greenland, on Titan)



*See talk by Jerry Watkins.

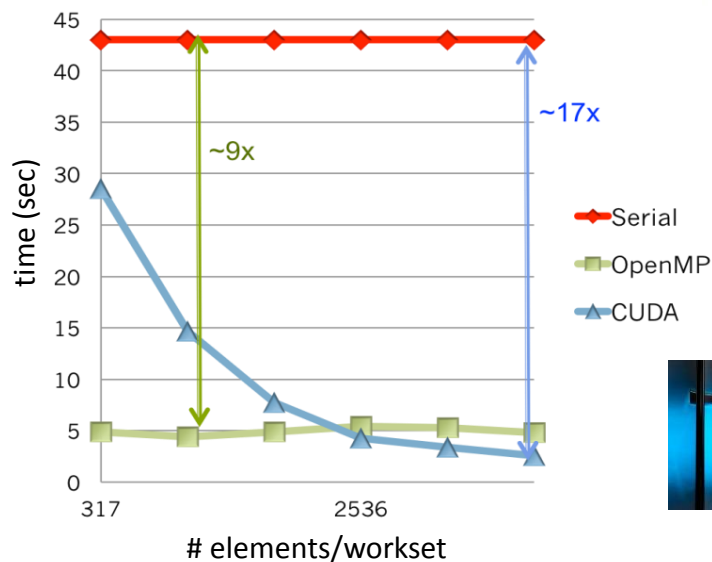
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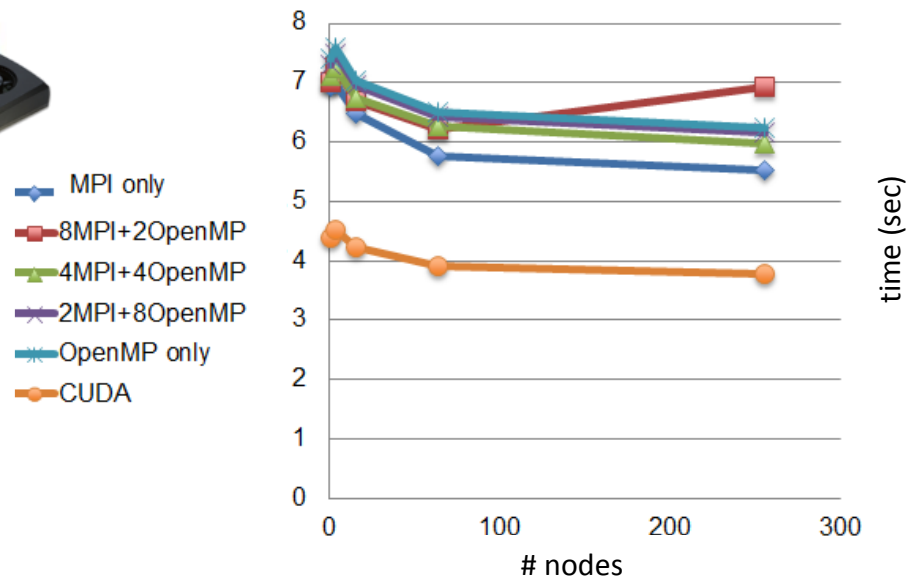
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FEA Time (Greenland, on Shannon)



Compute Time (Greenland, on Titan)



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Deterministic Inversion



Deterministic Inversion: Estimation of Ice Sheet Initial State

Objective: find ice sheet initial state that

- Matches observations (e.g., surf. vel., temp., etc.)
- Matches present-day geometry (elevation, thickness).
- Is in “equilibrium” with climate forcings (SMB).

Unknown/uncertain variables:

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First-Order Stokes PDE Constrained Optimization Problem:

$\min J(\beta, H)$ s.t. FO Stokes PDEs where

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- **Albany/FELIX** (FE assembly)
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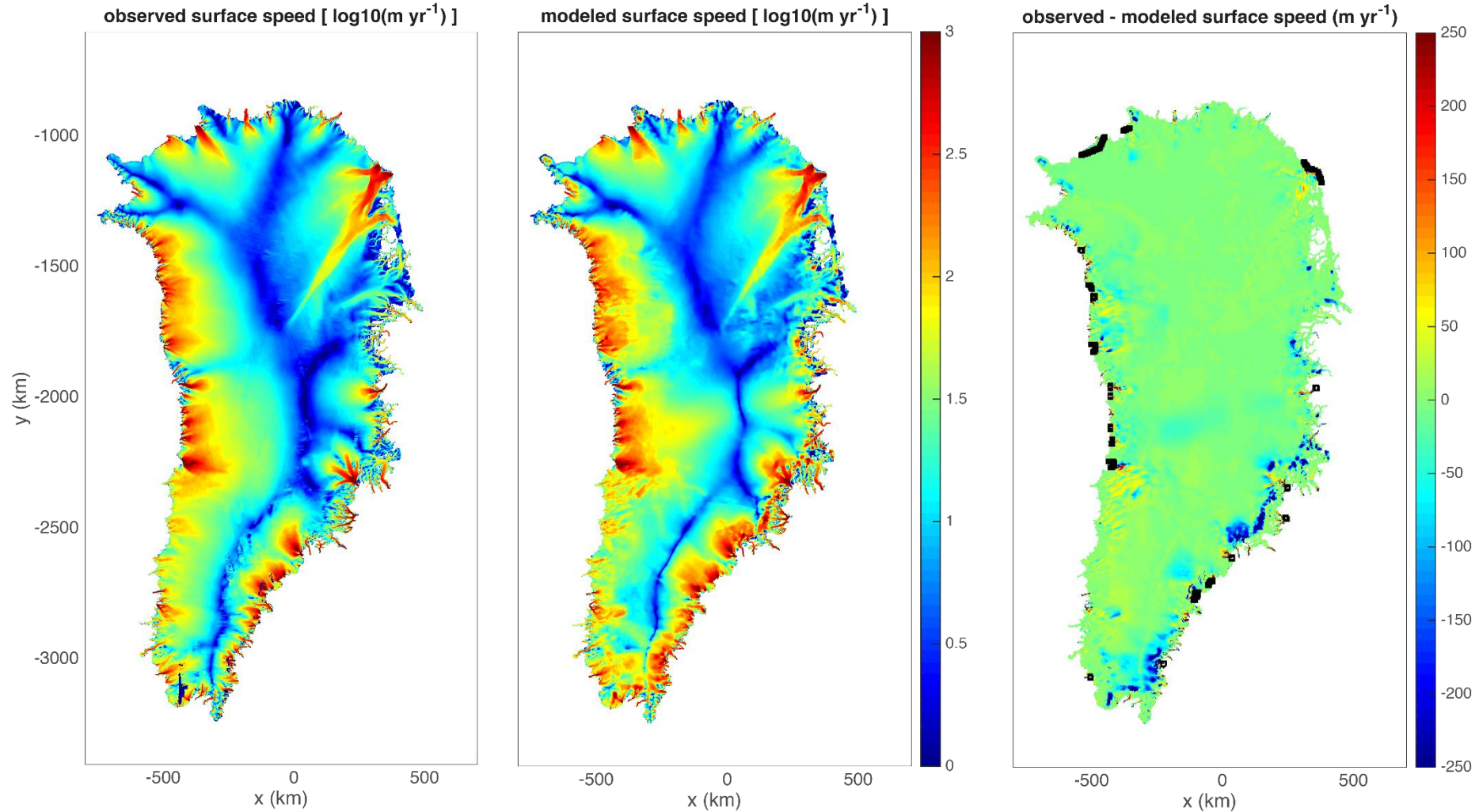
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→ significantly reduces non-physical model transients

Deterministic Inversion: 1km Greenland

Initial Condition*



*This initial condition was used for validation study, discussed later in the talk.

Uncertainty Quantification



Uncertainty Quantification Workflow

Goal: Uncertainty Quantification in 21st century sea level (QoI)

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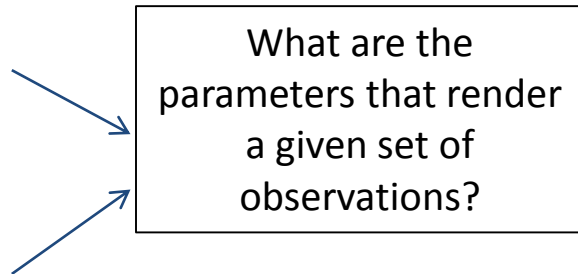
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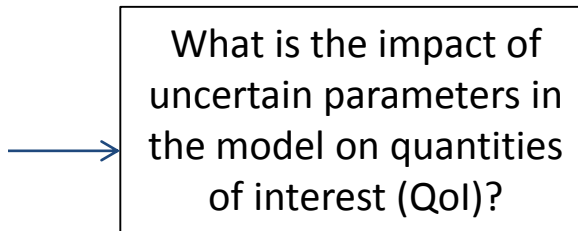
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What is the impact of uncertain parameters in the model on quantities of interest (QoI)?

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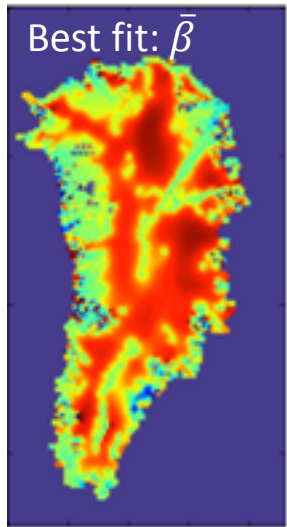
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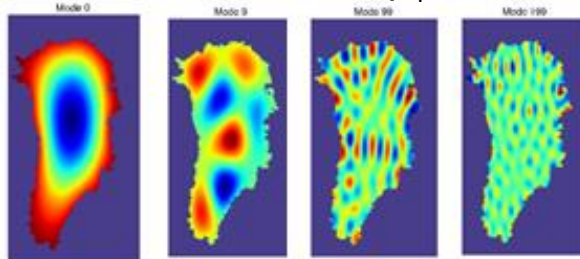
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$$\beta = \bar{\beta} + \sum_i \alpha_i \beta_i$$

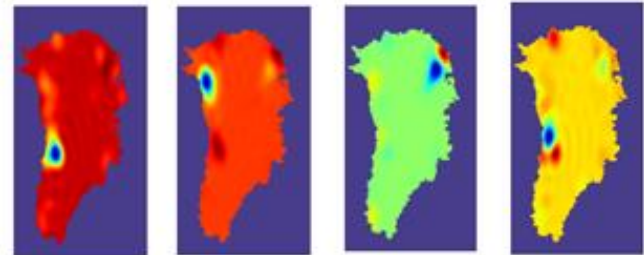
α_i : random samples
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Basis perturbations: β_i



Data-informed directions



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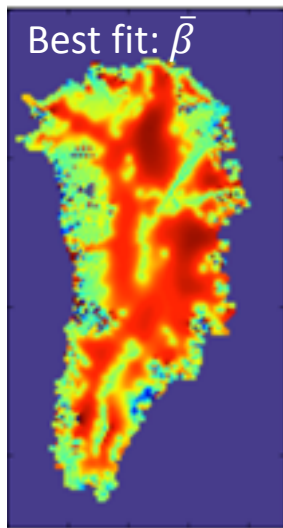
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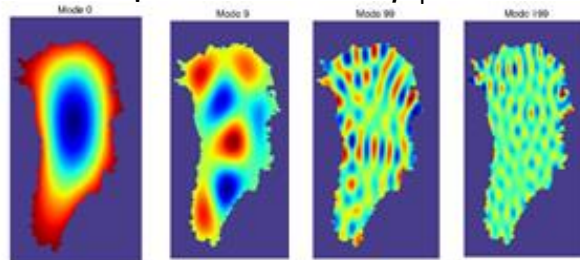
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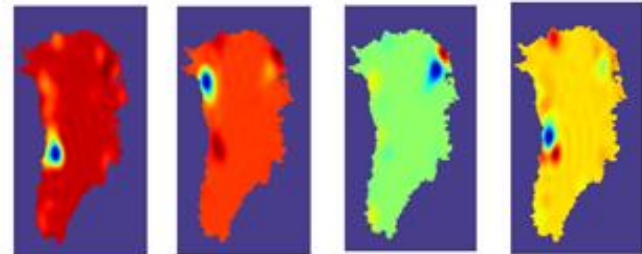
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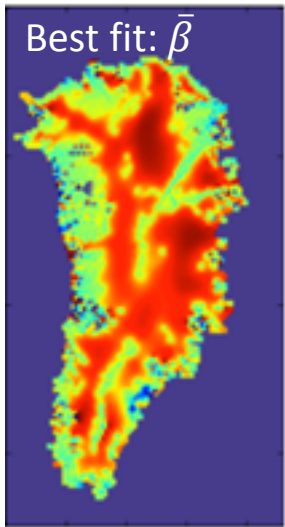


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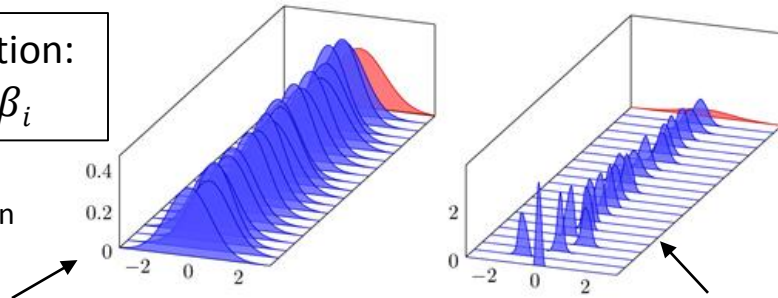
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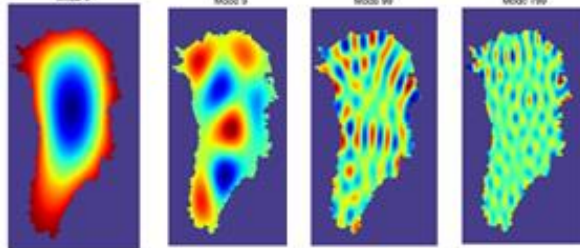
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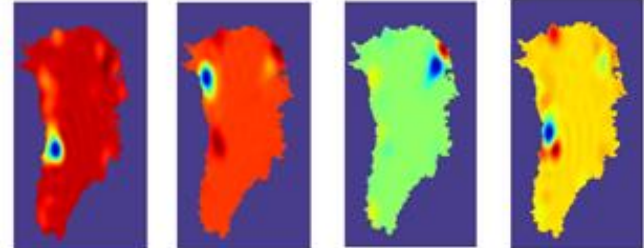


Marginal distributions of Gaussian posterior

Basis perturbations: β_i



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Forward Propagation

Propagate distribution obtained in Bayesian calibration through the model to get distributions on ***total ice mass loss/gain during 21st century***

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Forward Propagation

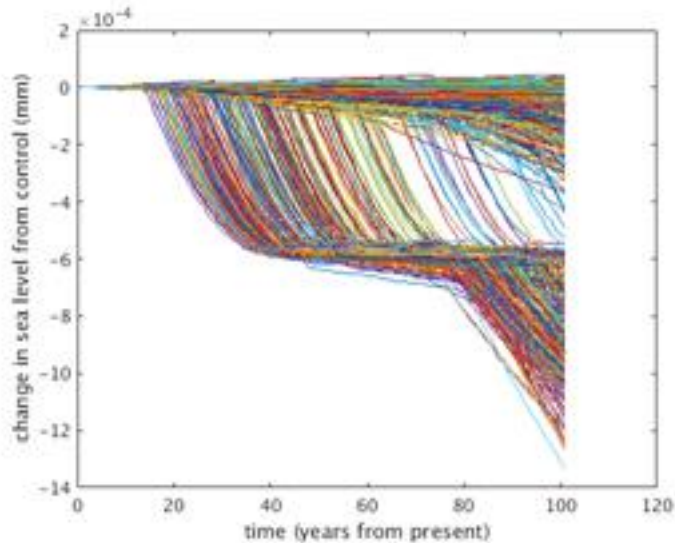
Propagate distribution obtained in Bayesian calibration through the model to get distributions on ***total ice mass loss/gain during 21st century***

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- Parameter distribution is the result of ***Bayesian calibration***.
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Sea level time-history for 1000 50-year forward runs with steady state forcing



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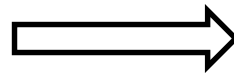
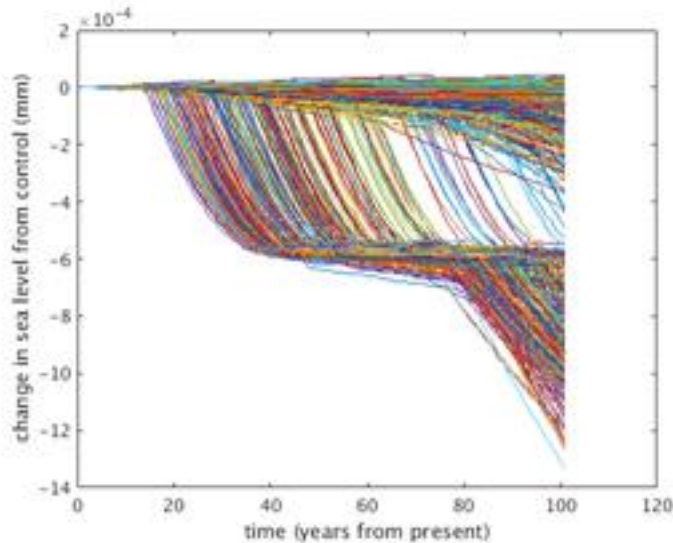
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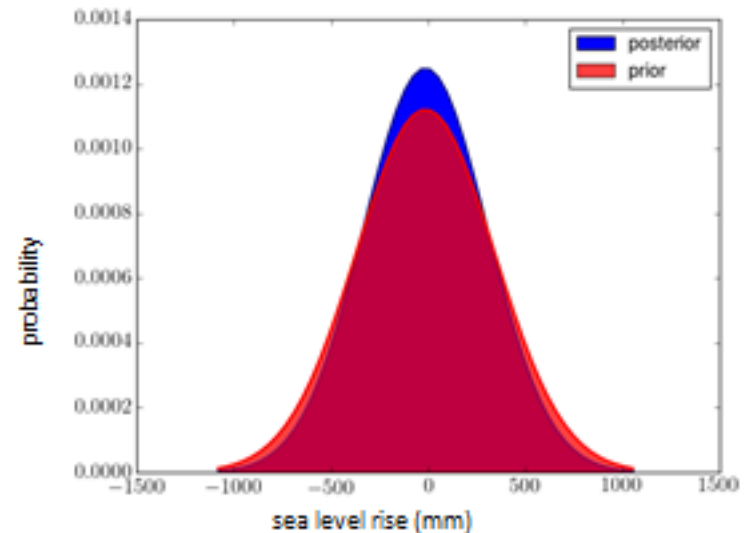
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- Use MCMC and emulator to perform ***uncertainty propagation*** [QUESO].



Sea level time-history for 1000 50-year forward runs with steady state forcing



PDFs of SLR



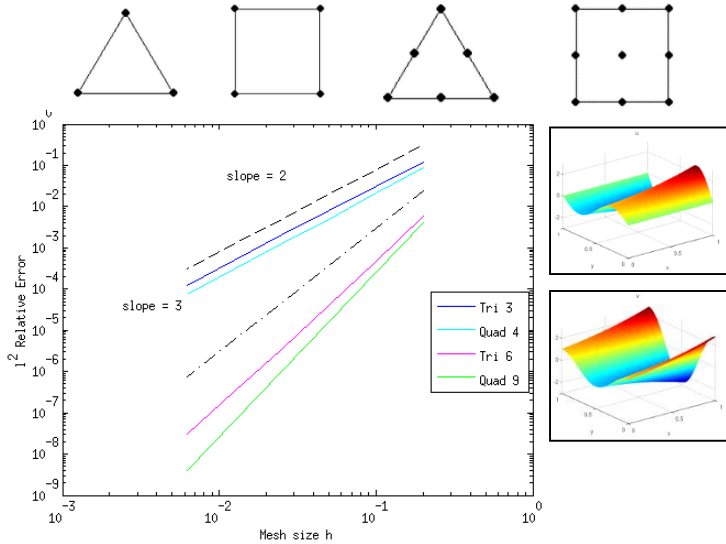
Verification and Validation



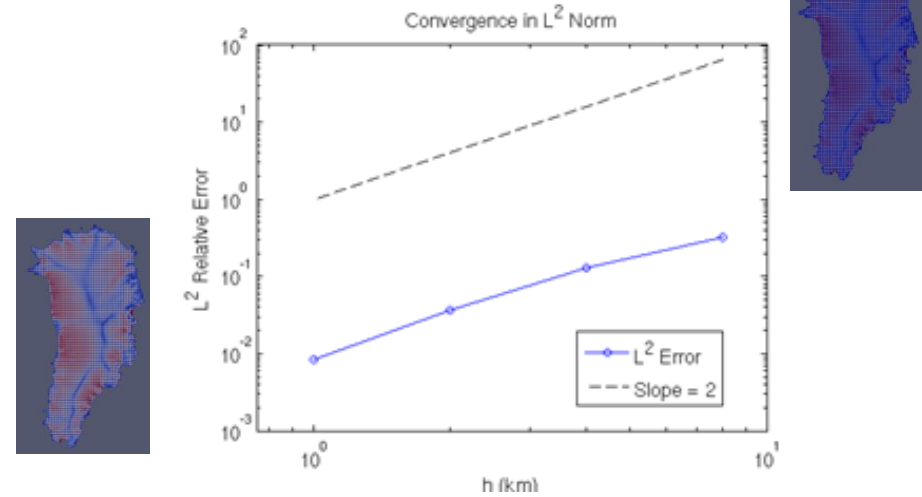
Verification

Is the code bug free?

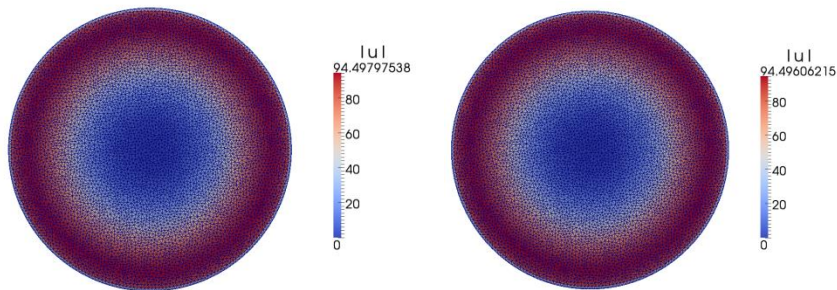
Stage 1: solution verification on 2D MMS problems.



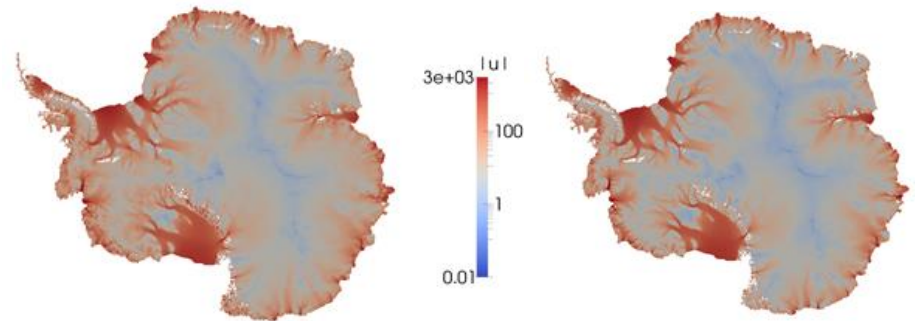
Stage 3: full 3D mesh convergence study on Greenland w.r.t. reference solution.



Stage 2: code-to-code comparisons on canonical ice sheet benchmarks (*Albany/FELIX* – left; *LifeV* – right).



Stage 4: reasonable solutions for large-scale realistic GIS & AIS problems (*Albany/FELIX* – left; reference solution – right).



Validation: Definition & Workflow

Validation: how well does model represent the real ice sheet?

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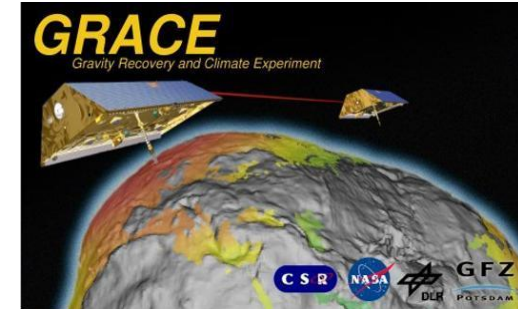
- There are currently (up to) 2 decades of large-scale **satellite** observations of Greenland ice sheet geometry change:

ICESat1	2003 – 2009
GRACE	2002 – 201? (ongoing)

Validation Workflow (with LANL & NASA):

- Run *CISM-Albany* for period where observations exist.
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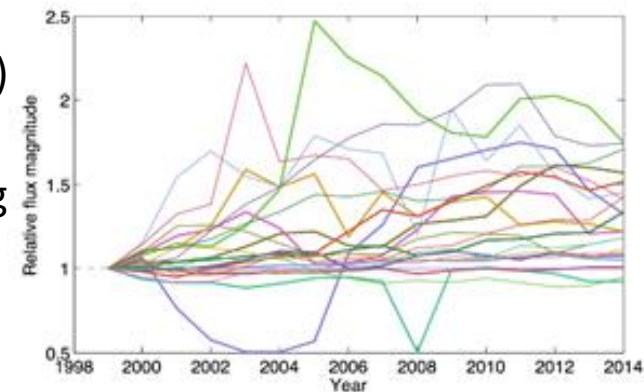
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- **Model forcing:** monthly surface mass balance (SMB) anomalies from RACMO2 and/or outlet glacier flux-forcing (FF) at grounding line applied.

Validation: Definition & Workflow

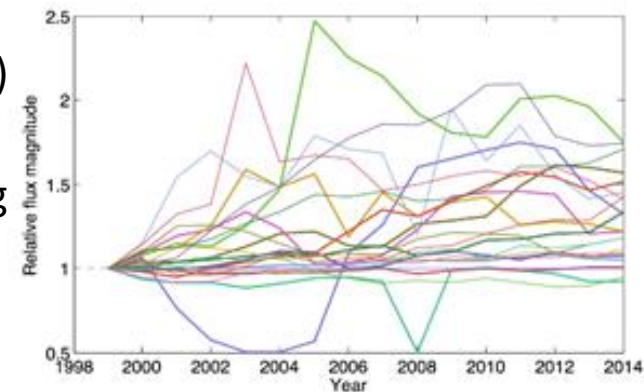
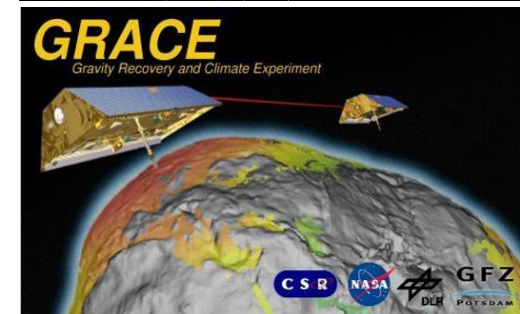
Validation: how well does model represent the real ice sheet?

- There are currently (up to) 2 decades of large-scale **satellite** observations of Greenland ice sheet geometry change:

ICESat1	2003 – 2009
GRACE	2002 – 201? (ongoing)

Validation Workflow (with LANL & NASA):

- Run *CISM-Albany* for period where observations exist.
- Process model output and observations for comparison.
- Evaluate model performance relative to observations.
 - **ICESat**: ice sheet surface elevation [state comparison]
 - **GRACE**: rate of mass change [trend comparison]

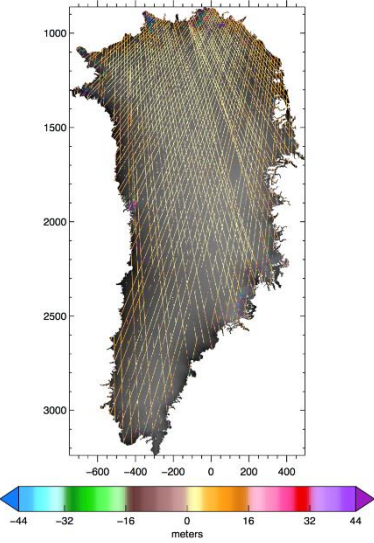


- **Validation time periods:** 2003-2009 (IceSAT), 2003-2011 (GRACE)
- **Model forcing:** monthly surface mass balance (SMB) anomalies from RACMO2 and/or outlet glacier flux-forcing (FF) at grounding line applied.
- **Initial condition** (1km GIS) obtained through deterministic inversion (shown on earlier slide).

Validation Results

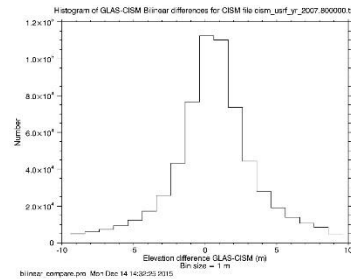
Surface Elevation Comparison [IceSAT]

Oct. 07 bilinear differences

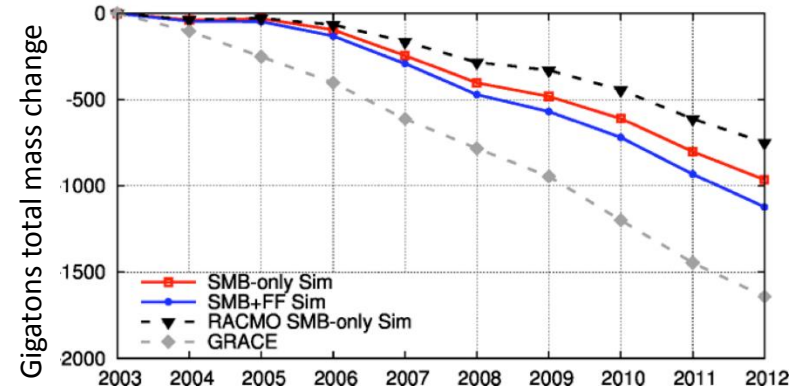


Surface elevation predictions (states) agree well with **GLAS** (*Geoscience Laser Altimeter System aboard ICESat*): **mean differences are <1 m**

Histogram of Oct. 07 bilinear differences



Whole Ice Sheet Mass Trends [GRACE]

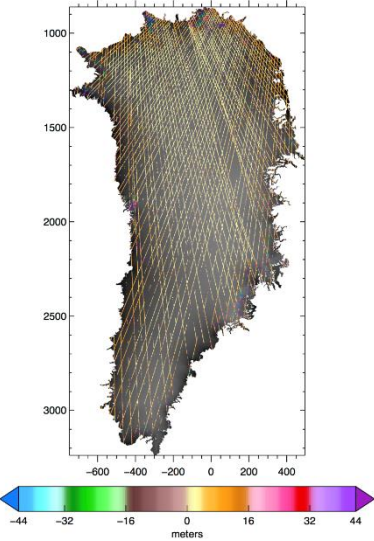


CISM-Albany results (blue, red lines) **closer to observations** (gray line) than older “idealized” simulations (black line).

Validation Results

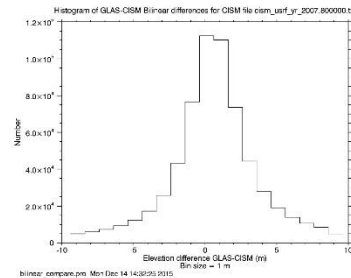
Surface Elevation Comparison [IceSAT]

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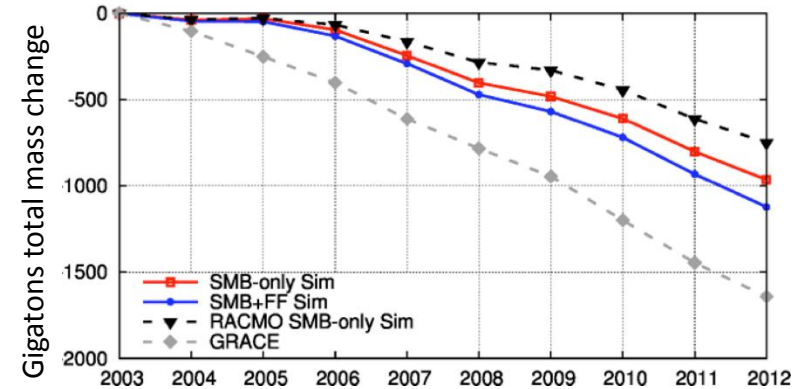


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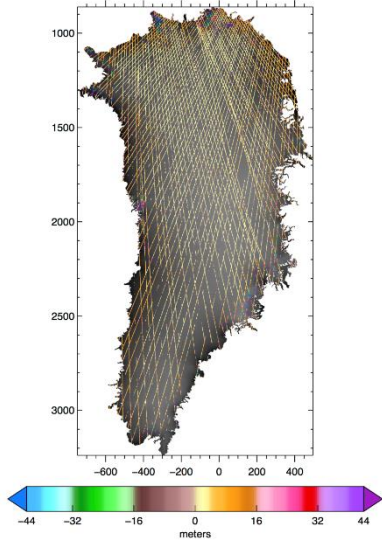
Main Takeaway from Validation Study

Current generation ice sheet models, when appropriately forced, show **skill** at mimicking ice sheet observations

Validation Results

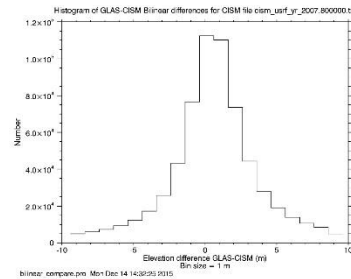
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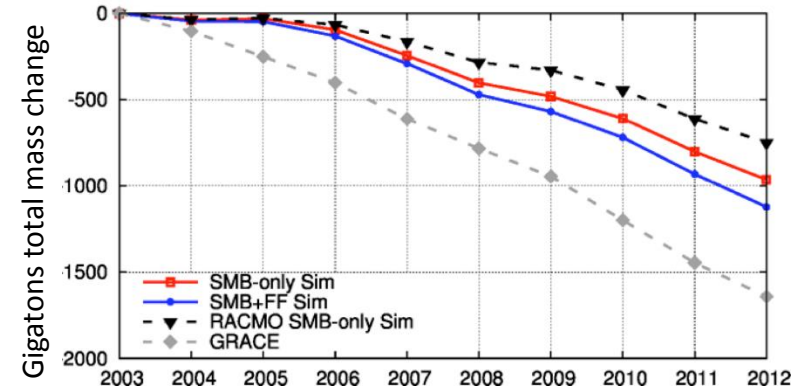


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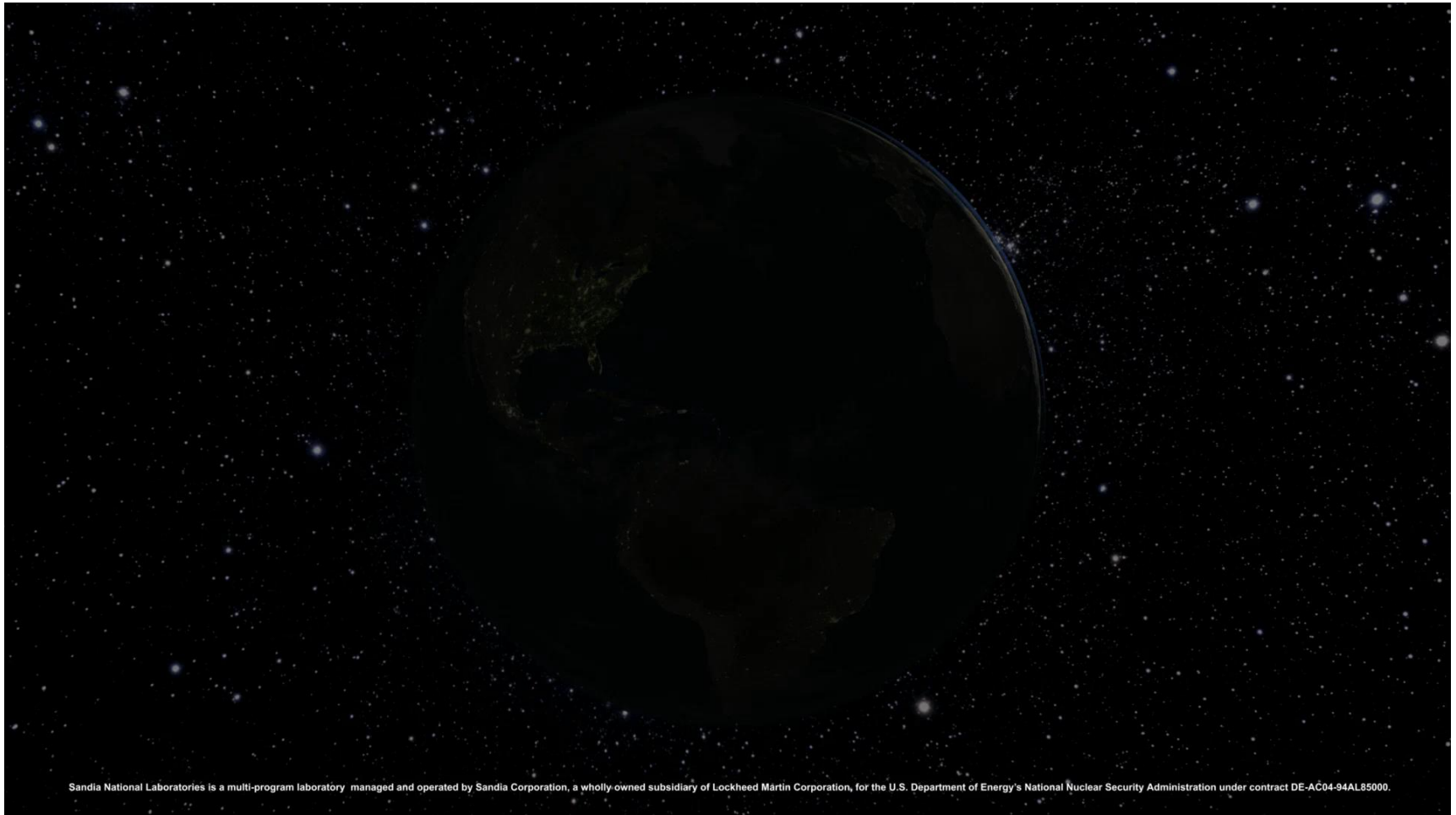
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Main Takeaway from Validation Study

Current generation ice sheet models, **when appropriately forced**, show **skill** at mimicking ice sheet observations

- **Clear improvement over a decade ago:** SLR projections from ice sheet models were not included in the IPCC’s AR4 b/c models could not explain observed ice dynamical behaviors.

Cool Movie of Validation Results



Video acknowledgement: B. Carvey (SNL)

Summary and Future Work

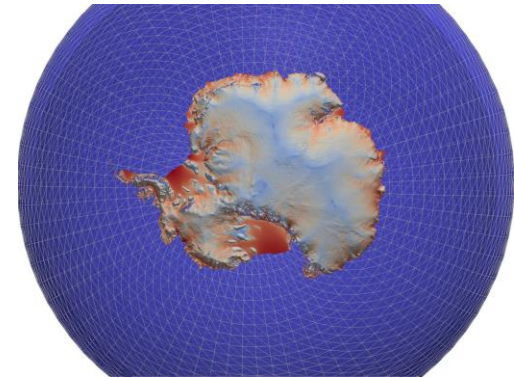


Summary:

- We have developed the *Albany/FELIX* land-ice solver, which is:
 - **Scalable, fast, robust.**
 - Coupled to *CISM* and *MPAS* codes for **dynamic** runs and integration into ESMs.
 - **Verified** and **validated.**
 - **Portable** to new and emerging architecture machines.
 - Equipped with **advanced analysis capabilities** (deterministic inversion, UQ).
 - A **production code** in Albany.

Albany/FELIX work not covered in this talk:

- **Semi-implicit** FO Stokes-thickness coupling methods.
- **Temperature solver** in *Albany/FELIX*.
- More sophisticated **basal hydrology models.**
- FO Stokes model on **spherical grids** via stereographic projection.



Ongoing/future work:

- **Science runs** using *CISM-Albany* and *MPAS-Albany*.
- Code optimizations for **new architecture machines** (GPUs, Intel Xeon Phis).
- Improving **UQ** workflow / algorithms, towards paper.
- **Proposal** for follow-up funding (SciDaC4).
- Delivering code to climate community and **coupling to ESMs.**

Funding/Acknowledgements

Support for this work was provided through Scientific Discovery through Advanced Computing (**SciDAC**) projects funded by the U.S. Department of Energy, Office of Science (**OSCR**), Advanced Scientific Computing Research and Biological and Environmental Research (**BER**) → **PISCEES SciDAC Application Partnership**.



PISCEES team members: K. Evans, M. Gunzburger, M. Hoffman, C. Jackson, P. Jones, W. Lipscomb, M. Perego, S. Price, A. Salinger, I. Tezaur, R. Tuminaro, P. Worley.

Trilinos/DAKOTA collaborators: M. Eldred, J. Jakeman, E. Phipps, L. Swiler.

Computing resources: NERSC, OLCF.

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Stokes Ice Flow Equations

Ice behaves like a very viscous shear-thinning fluid (similar to lava flow) and is modeled using nonlinear incompressible Stokes' equations.

- Nonlinear incompressible Stokes' ice flow equations (momentum balance):

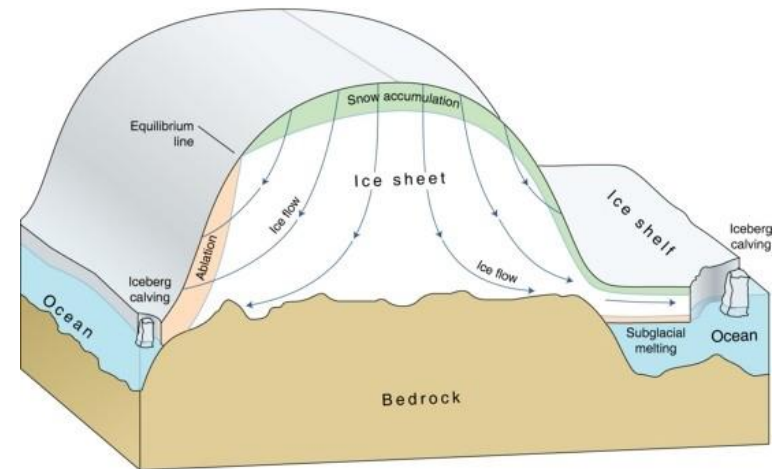
$$\begin{cases} -\nabla \cdot \boldsymbol{\sigma} = \rho \mathbf{g} \\ -\nabla \cdot \mathbf{u} = 0 \end{cases}, \quad \text{in } \Omega$$

with

$$\boldsymbol{\sigma} = 2\mu \dot{\boldsymbol{\epsilon}} - p\mathbf{I}, \quad \epsilon_{ij} = \frac{1}{2} \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right)$$

and nonlinear "**Glen's law**" viscosity

$$\mu = \frac{1}{2} A(T)^{-\frac{1}{n}} \left(\frac{1}{2} \sum_{ij} \epsilon_{ij}^2 \right)^{\left(\frac{1}{2n} - \frac{1}{2} \right)}, \quad n = 3.$$



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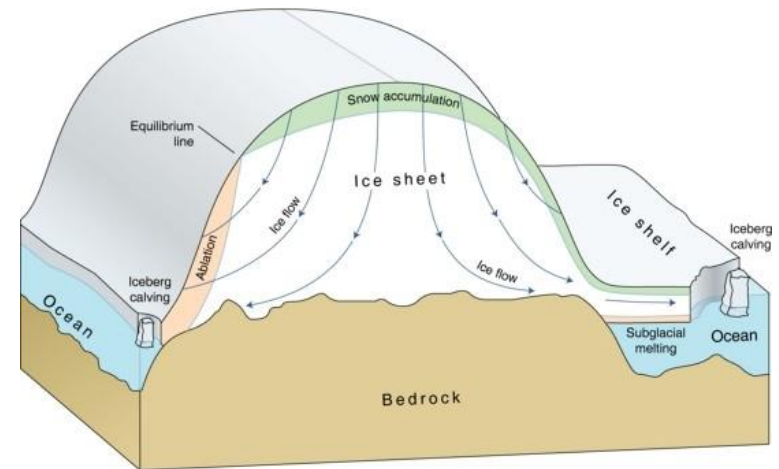
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→ "nasty" saddle point problem