Albany/FELIX: A Robust & Scalable Trilinos-Based Finite-Element Ice Flow Dycore Built for Advanced Architectures & Analysis

I. Tezaur<sup>1</sup>, A. Salinger<sup>1</sup>, M. Perego<sup>1</sup>, J. Jakeman<sup>1</sup>, M. Eldred<sup>1</sup>,
 I. Demeshko<sup>1</sup>, R. Tuminaro<sup>1</sup>, S. Price<sup>2</sup>

<sup>1</sup> Sandia National Laboratories Livermore, CA and Albuquerque, NM, USA <sup>2</sup> Los Alamos National Laboratory Los Alamos, NM, USA



1

8<sup>th</sup> International Congress on Industrial and Applied Mathematics (ICIAM 2015) August 10-14, 2015 Beijing, China



Sandia National Laboratories

SAND2015-6419C

### Outline

*Albany/FELIX* = new land-ice solver with *next-generation* capabilities.

- Overview:
  - First Order (FO) Stokes model.
  - Albany/FELIX First-Order (FO) Stokes diagnostic solver.
  - CISM-Albany and MPAS-Albany codes for prognostic simulations of the ice sheet evolution.
- Uncertainty Quantification (UQ):
  - Deterministic inversion.
  - Bayesian calibration.
  - Forward propagation of uncertainty.
- Performance portability.
- **Summary** and ongoing work.
- Questions?





#### Outline

*Albany/FELIX* = new land-ice solver with *next-generation* capabilities.

• Overview:

- First Order (FO) Stokes model.
- Albany/FELIX First-Order (FO) Stokes diagnostic solver.
- CISM-Albany and MPAS-Albany codes for prognostic simulations of the ice sheet evolution.
- Uncertainty Quantification (UQ):
  - Deterministic inversion.
  - Bayesian calibration.
  - Forward propagation of uncertainty.
- Performance portability.
- **Summary** and ongoing work.
- Questions?







# The PISCEES Project and the *Albany/FELIX* Solver



This

talk

"PISCEES" = Predicting Ice Sheet Climate & Evolution at Extreme Scales 5 year project funded by SciDAC, which began in June 2012

# <u>Sandia's Role in the PISCEES Project:</u> to develop and support a robust and scalable land ice solver based on the "First-Order" (FO) Stokes approximation

- Steady-state stress-velocity solver based on FO Stokes physics is known as Albany/FELIX\*.
- <u>Requirements for Albany/FELIX:</u>
  - Scalable, fast, robust.

Dycore will provide actionable predictions of 21<sup>st</sup> century sea-level rise (including uncertainty).

- Dynamical core (dycore) when coupled to codes that solve thickness and temperature evolution equations (*CISM/MPAS LI* codes).
- Advanced analysis capabilities (adjoint-based deterministic inversion, Bayesian calibration, UQ, sensitivity analysis).
- Performance-portability.

Albany/FELIX Solver (steady): Ice Sheet PDEs (First Order Stokes) (stress-velocity solve)



\*FELIX="Finite Elements for Land Ice eXperiments"

#### CISM/MPAS Land Ice Codes (dynamic):

Ice Sheet Evolution PDEs

(thickness, temperature evolution)



5

# The First-Order Stokes Model for Ice Sheets & Glaciers

Ice sheet dynamics are given by the *"First-Order" Stokes PDEs*: approximation\* to viscous incompressible *quasi-static* Stokes flow with power-law viscosity.

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

• Viscosity  $\mu$  is nonlinear function given by "*Glen's law"*:

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

- Relevant boundary conditions:
  - Stress-free BC:  $2\mu \dot{\boldsymbol{\epsilon}}_i \cdot \boldsymbol{n} = 0$ , on  $\Gamma_s$
  - Floating ice BC:

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

• **Basal sliding BC:**  $2\mu \dot{\epsilon}_i \cdot n + \beta u_i = 0$ , on  $\Gamma_\beta$ 









# Algorithmic Choices for Albany/FELIX: Discretization & Meshes

- **Discretization:** unstructured grid finite element method (FEM)
  - Can handle readily complex geometries.

- Natural treatment of stress boundary conditions.
- Enables regional refinement/unstructured meshes.
- Wealth of software and algorithms.
- Meshes: can use any mesh but interested specifically in
  - Structured hexahedral meshes (compatible with CISM).
  - Tetrahedral meshes (compatible with MPAS LI)
  - **Unstructured Delaunay triangle** meshes with regional refinement based on gradient of surface velocity.
  - All meshes are extruded (structured) in vertical direction as tetrahedra or hexahedra.

# Algorithmic Choices for Albany/FELIX: Nonlinear & Linear Solver

- **Nonlinear solver:** full Newton with analytic (automatic differentiation) derivatives and homotopy continuation
  - Most robust and efficient for steady-state solves.
  - Jacobian available for preconditioners and matrix-vector products.
  - Analytic sensitivity analysis.

- Analytic gradients for inversion.
- Linear solver: preconditioned iterative method
  - Solvers: Conjugate Gradient (CG) or GMRES
  - Preconditioners: ILU or algebraic multi-grid (AMG)



# The Albany/FELIX Solver: Implementation in Albany using Trilinos

8



Use of **Trilinos** components has enabled the **rapid** development of the **Albany/FELIX** First Order Stokes dycore!



### The Albany/FELIX Solver is Verified, Scalable, Fast and Robust!





**Ice Sheet Evolution Models** 

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

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

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

• Temperature equation (advection-diffusion):

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

(energy balance).

10

- 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  $(H > H_{min})$ 



Ice Sheet Evolution Models

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

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

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

• Temperature equation (advection-diffusion):

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

(energy balance).

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









Ice Sheet Evolution Models

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

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

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

• Temperature equation (advection-diffusion):

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

(energy balance).

12

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





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



# Interfaces to CISM and MPAS LI for Transient Simulations



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 LI)



#### Outline

*Albany/FELIX* = new land-ice solver with *next-generation* capabilities.

• Overview:

- First Order (FO) Stokes model.
- Albany/FELIX First-Order (FO) Stokes diagnostic solver.
- **CISM-Albany** and **MPAS-Albany** codes for prognostic simulations of the ice sheet evolution.
- Uncertainty Quantification (UQ):
  - Deterministic inversion.
  - Bayesian calibration.
  - Forward propagation of uncertainty.
- Performance portability.
- **Summary** and ongoing work.
- Questions?





#### Uncertainty Quantification (UQ) Problem Definition

Quantity of Interest (QoI) in Ice Sheet Modeling: total ice mass loss/gain during  $21^{st}$  century  $\rightarrow$  sea level rise prediction.

There are several sources of uncertainty, most notably:

- Climate forcings (e.g., surface mass balance).
- Basal friction ( $\beta$ )

15

- Bedrock topography
- Geothermal heat flux
- Model parameters (e.g., Glen's flow law exponent)



Basal sliding BC:  $2\mu \dot{\boldsymbol{\epsilon}}_i \cdot \boldsymbol{n} + \beta u_i = 0$ , on  $\Gamma_{\beta}$ 



#### Uncertainty Quantification (UQ) Problem Definition

Quantity of Interest (QoI) in Ice Sheet Modeling: total ice mass loss/gain during  $21^{st}$  century  $\rightarrow$  sea level rise prediction.

There are several sources of uncertainty, most notably:

- Climate forcings (e.g., surface mass balance).
- Basal friction ( $\beta$ ).

16

- Bedrock topography.
- Geothermal heat flux.
- Model parameters (e.g., Glen's flow law exponent).

As a first step, we focus on effect of uncertainty in **basal friction** ( $\beta$ ) only.



Basal sliding BC:  $2\mu \dot{\boldsymbol{\epsilon}}_i \cdot \boldsymbol{n} + \beta u_i = 0$ , on  $\Gamma_{\beta}$ 





### Uncertainty Quantification Workflow

**Goal:** Uncertainty Quantification in 21<sup>st</sup> century sea level (QoI)

- Deterministic inversion: perform adjoint-based deterministic inversion to estimate initial ice sheet state (i.e., characterize the present state of the ice sheet to be used for performing prediction runs).
- Bayesian calibration: construct the posterior distribution using Markov Chain Monte Carlo (MCMC) run on an emulator of the forward model.
- Forward propagation: sample the obtained distribution and perform ensemble of forward propagation runs to compute the uncertainty in the QoI.

What are the parameters that render a given set of observations?

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



# Deterministic Inversion: Estimation of Ice Sheet Initial State

Objective: find ice sheet initial state that

- Matches observations (e.g., surface velocity, temperature, etc.)
- Matches present-day geometry (elevation, thickness).
- Is in "equilibrium" with climate forcings (SMB).

**Approach:** invert for unknown/uncertain ice sheet model parameters.

• Significantly reduces non-physical transients without model spin-up.

#### Available data/measurements:

- Ice extent and surface topography.
- Surface velocity.
- Surface mass balance (SMB).
- Ice thickness *H* (sparse measurements).

#### Field to be estimated:

- Basal friction  $\beta$  (spatially variable proxy for all basal processes).
- Ice thickness H (allowed to be weighted by observational uncertainties).

#### Assumptions:

- Ice flow described by FO Stokes equations.
- Ice is close to mechanical equilibrium.
- Temperature field is given.





### **Deterministic Inversion:** Greenland

**First Order Stokes PDE Constrained Optimization Problem:** 

$$J(\beta, H) = \frac{1}{2}\alpha_{v}\int_{\Gamma_{top}}|\boldsymbol{u} - \boldsymbol{u}^{obs}|^{2}ds + \frac{1}{2}\alpha\int_{\Gamma}|div(\boldsymbol{U}H) - SMB|^{2}ds + \frac{1}{2}\alpha_{H}\int_{\Gamma_{top}}|H - H^{obs}|^{2}ds + \mathcal{R}(\beta) + \mathcal{R}(H)$$

- Minimize difference between:
  - Computed and measured *surface velocity*  $(u^{obs}) \rightarrow common$
  - Computed divergence flux and measured surface mass **balance (SMB)** → novel
  - Computed and *reference thickness* (*H*<sup>obs</sup>) → novel
- Control variables:

19

- Basal friction ( $\beta$ ).
- Thickness (H).

#### Software for adjoint-based inversion:

- Albany/FELIX (assembly)
- *Trilinos* (linear/nonlinear solvers)
- **ROL** (gradient-based optimization)
  - Limited memory BFGS.
  - Backtrack line-search.







Estimated (left) vs. reference surface velocity (right)



# Deterministic Inversion: Antarctica (basal friction only)

FO Stokes PDE Constrained Optimization Problem:

20

 $J(\beta) = \frac{1}{2} \int_{\Gamma_{top}} \alpha |\boldsymbol{u} - \boldsymbol{u}^{obs}|^2 ds + \mathcal{R}(\beta)$ 

*Geometry:* Cornford, Martin *et al.* (in prep.) *Bedmap2:* Fretwell et al., 2013 *Temperature:* Pattyn, 2010.



Antarctic ice sheet inversion performed on **700K** parameters



Albany/FELIX has been hooked up to **DAKOTA** (in "black-box" mode) for **UQ**/ **Bayesian calibration.** 

> **Difficulty in UQ**: "Curse of Dimensionality" The  $\beta$ -field inversion problem has O(100K) dimensions!



Albany/FELIX has been hooked up to **DAKOTA** (in "black-box" mode) for **UQ**/ **Bayesian calibration.** 

22

**Difficulty in UQ**: "Curse of Dimensionality" The  $\beta$ -field inversion problem has O(100K) dimensions!

**<u>Approach</u>**: Reduce O(100K) dimensional problem to O(10) dimensional problem.



Albany/FELIX has been hooked up to **DAKOTA** (in "black-box" mode) for **UQ**/ **Bayesian calibration.** 

> **Difficulty in UQ**: "Curse of Dimensionality" The  $\beta$ -field inversion problem has O(100K) dimensions!

**Approach:** Reduce O(100K) dimensional problem to O(10) dimensional problem.

• For initial demonstration of workflow, we use the *Karhunen-Loeve Expansion (KLE)*:



*Albany/FELIX* has been hooked up to **DAKOTA** (in "black-box" mode) for **UQ/ Bayesian calibration.** 

**Difficulty in UQ**: "Curse of Dimensionality" The  $\beta$ -field inversion problem has O(100K) dimensions!

**Approach:** Reduce O(100K) dimensional problem to O(10) dimensional problem.

- For initial demonstration of workflow, we use the *Karhunen-Loeve Expansion (KLE)*:
  - 1. Assume analytic covariance kernel  $C(r_1, r_2) = exp\left(-\frac{(r_1 r_2)^2}{L^2}\right)$ .



*Albany/FELIX* has been hooked up to **DAKOTA** (in "black-box" mode) for **UQ/ Bayesian calibration.** 

**Difficulty in UQ**: "Curse of Dimensionality" The  $\beta$ -field inversion problem has O(100K) dimensions!

**Approach:** Reduce O(100K) dimensional problem to O(10) dimensional problem.

- For initial demonstration of workflow, we use the *Karhunen-Loeve Expansion (KLE)*:
  - 1. Assume analytic covariance kernel  $C(r_1, r_2) = exp\left(-\frac{(r_1 r_2)^2}{L^2}\right)$ .
  - 2. Perform eigenvalue decomposition of *C*.



*Albany/FELIX* has been hooked up to **DAKOTA** (in "black-box" mode) for **UQ/ Bayesian calibration.** 

**Difficulty in UQ**: "Curse of Dimensionality" The  $\beta$ -field inversion problem has O(100K) dimensions!

**Approach:** Reduce O(100K) dimensional problem to O(10) dimensional problem.

- For initial demonstration of workflow, we use the *Karhunen-Loeve Expansion (KLE)*:
  - 1. Assume analytic covariance kernel  $C(r_1, r_2) = exp\left(-\frac{(r_1 r_2)^2}{L^2}\right)$ .
  - 2. Perform eigenvalue decomposition of *C*.
  - 3. Expand\*  $\beta \overline{\beta}$  in basis of eigenvectors  $\{\phi_k\}$  of *C*, with random variables  $\{\xi_k\}$ :

$$\beta(\omega) = \bar{\beta} + \sum_{k=1}^{K} \sqrt{\lambda_k} \boldsymbol{\phi}_k \xi_k(\omega)$$

 $\bar{\beta}$  = initial condition for  $\beta$ (result of deterministic inversion or spin-up)

Albany/FELIX has been hooked up to DAKOTA (in "black-box" mode) for UQ/ **Bayesian calibration.** 

> **Difficulty in UQ**: "Curse of Dimensionality" The  $\beta$ -field inversion problem has O(100K) dimensions!

**Approach:** Reduce O(100K) dimensional problem to O(10) dimensional problem.

- For initial demonstration of workflow, we use the *Karhunen-Loeve Expansion (KLE*):
- Offline  $\begin{bmatrix} 1. & \text{Assume analytic covariance kernel } C(r_1, r_2) = exp\left(-\frac{(r_1 r_2)^2}{L^2}\right). \\ 2. & \text{Perform eigenvalue decomposition of } C. \end{bmatrix}$

27

- Expand\*  $\beta \overline{\beta}$  in basis of eigenvectors  $\{\phi_k\}$  of *C*, with random variables  $\{\xi_k\}$ : 3.

$$\beta(\omega) = \bar{\beta} + \sum_{k=1}^{K} \sqrt{\lambda_k} \boldsymbol{\phi}_k \xi_k(\omega)$$

 $\bar{\beta}$  = initial condition for  $\beta$ (result of deterministic inversion or spin-up)



\*In practice, expansion is done on  $\log(\beta)$  to avoid negative values of  $\beta$ .

Albany/FELIX has been hooked up to **DAKOTA** (in "black-box" mode) for **UQ**/ **Bayesian calibration.** 

> **Difficulty in UQ**: "Curse of Dimensionality" The  $\beta$ -field inversion problem has O(100K) dimensions!

**Approach:** Reduce O(100K) dimensional problem to O(10) dimensional problem.

- For initial demonstration of workflow, we use the *Karhunen-Loeve Expansion (KLE*):
- Offline  $\begin{bmatrix} 1. & \text{Assume analytic covariance kernel } C(r_1, r_2) = exp\left(-\frac{(r_1 r_2)^2}{L^2}\right). \\ 2. & \text{Perform eigenvalue decomposition of } C. \end{bmatrix}$ 

  - Expand\*  $\beta \overline{\beta}$  in basis of eigenvectors  $\{\phi_k\}$  of *C*, with random variables  $\{\xi_k\}$ : 3.

Online

$$\beta(\omega) = \bar{\beta} + \sum_{k=1}^{K} \sqrt{\lambda_k} \phi_k \xi_k(\omega)$$

 $\bar{\beta}$  = initial condition for  $\beta$ (result of deterministic inversion or spin-up)

\*In practice, expansion is done on  $\log(\beta)$  to avoid negative values of  $\beta$ .

Inference/calibration is for coefficients of KLE  $\Rightarrow$  significant dimension reduction.



**Disclaimer**: results presented demonstrate that we have UQ workflow in place; quantifying uncertainty in  $\beta$  and SLR will require re-running with better data.

29

### Bayesian Calibration: Illustration on 4km GIS Problem

- Mean  $\overline{\beta}$  field obtained through spin-up over 100 years (cheaper than inversion, gives reasonable agreement with present-day velocity field).
- Correlation length *L* selected s.t. slow decay of KLE eigenvalues to enable refinement *(left):* 10 KLE modes capture 27.3% of covariance energy.



• Mismatch function (calculated in *Albany/FELIX*):











- PCE emulator was formed for the mismatch  $J(\beta)$  using uniform [-1,1] prior distributions and 286\* high-fidelity runs on Hopper (*DAKOTA*).
- For calibration, MCMC was performed on the PCE with 2K samples (QUESO). \*286 points = 3<sup>rd</sup> degree polynomial in 10D



# Bayesian Calibration: Illustration on 4km GIS Problem (cont'd)

• *Posterior distributions* for 10 KLE coefficients:



- Distributions are peaked rather than uniform  $\Rightarrow$  data informed the posteriors.
- MAP point: ξ = (0.372, -0.679, -0.420, -0.189, -7.38e-2, -0.255, 0.449, -0.757, 0.847, -0.447)



# Bayesian Calibration: Illustration on 4km GIS Problem (cont'd)



- Ice is too fast at MAP point. Possible explanations:
  - Surrogate error (based on cross-validation).
  - Mean field error.

31

• Bad modes and/or not enough modes.

Mismatch  $J(\beta)$  at MAP point: 1.87 × mismatch at  $\overline{\beta}$ 



# Bayesian Calibration: Building Gaussian Distribution using Hessian

Sandia

National Laboratories

 Hessian of the merit functional (velocity mismatch) can provide a way to compute the covariance of a Gaussian posterior:

$$\boldsymbol{C}_{post} = (\boldsymbol{C}_{prior}\boldsymbol{H}_{misfit} + \boldsymbol{I})^{-1}\boldsymbol{C}_{prior}$$

• We want to limit only the most important directions (eigenvectors) of *C*<sub>post</sub>.

32



#### Issue: there are too many (~1000) significant eigenvalues

*Right*: log-linear plot of the spectra of a prior-preconditioned data misfit Hessian at the MAP point for two successively finer parameter/state meshes of the inverse ice sheet problem.



### **Forward Propagation**

PCE Emulator  
$$\beta(\omega) = \bar{\beta} + \sum_{k=1}^{K} \sqrt{\lambda_k} \phi_k \xi_k(\omega)$$

Model realizations Forward propagation (e.g., 2000-2050) DAKOTA, Albany/FELIX Qol( $\beta$ ) (total ice mass loss)

- Parameter ( $\beta$ ) distribution can either be assumed to be Gaussian (based on Hessian information) or can be the result of Bayesian calibration.
- Emulator is built using DAKOTA coupled with CISM-Albany for forward runs.
  - Use *compressed sensing* to adaptively select significant modes that affect QoI. The hope is that only a few modes affect the QoI.
  - Could use cheaper physical models to reduce computational time of forward model.



• MCMC (*QUESO*) used to perform uncertainty propagation.





34

**Disclaimer**: results presented demonstrate that we have UQ workflow in place; quantifying uncertainty in  $\beta$  and SLR will require re-running with better data.

# Forward Propagation: Illustration on 4km GIS Problem

#### Procedure:

• We first ran 66\* CISM-Albany high-fidelity simulations on Hopper with  $\beta$  sampled from a uniform [-1,1] distribution and no forcing for 50 years.



Left: SLR distribution from ensemble of 66 highfidelity simulations (differenced against control run using the  $\bar{\beta}$ distribution). All 66 runs ran to completion out-ofthe-box on Hopper!



Above:  $\beta$ , velocity and thickness perturbations. Ice thickness changed > 500m in some places.

- We then used the results of these runs to create a PCE emulator for the SLR.
- Using emulator, propagated posterior distributions computed in Bayesian calibration (using KLE) through the model to get posteriors on SLR (MCMC on PCE emulator with 2K samples).



# Forward Propagation: Illustration on 4km GIS Problem (cont'd)

**Expected (black):** normal distribution centered around 0 SLR since no forcing.

35

**Prior informed (green)**: uniform distribution translates to distribution skewed w.r.t. model outputs.

- Larger fraction of the ice sheet currently has a  $\beta$  value that forces no (or slow) basal sliding.
- Areas with little sliding: not affected by increase in  $\beta$ , but greatly affected by decrease in  $\beta$  (velocity in these regions will change significantly from initial condition).
- Since we sample from a uniform distribution when perturbing  $\beta$ , we expect to see a disproportionately large signal when reducing  $\beta$  vs. increasing it.



#### **Posterior informed (blue):** centered on positive tail of prior – not consistent with observations.

- Could be due to "ad hoc"  $\beta$  used as mean field (spin-up over 100 years).
- May be that emulator was been built with a (non-physical) positive mass balance while calibration was done on present-day observations (consistent with ice losing mass).



PDF of SLR

# Forward Propagation: Illustration on 4km GIS Problem (cont'd)

**Expected (black):** normal distribution centered around 0 SLR since no forcing.

**Prior informed (green)**: uniform distribution translates to distribution skewed w.r.t. model outputs.

- Larger fraction of the ice sheet currently has a  $\beta$  value that forces no (or slow) basal sliding.
- Areas with little sliding: not affected by increase in  $\beta$ , but greatly affected by decrease in  $\beta$  (velocity in these regions will change significantly from initial condition).
- Since we sample from a uniform distribution when perturbing  $\beta$ , we expect to see a disproportionately large signal when reducing  $\beta$  vs. increasing it.



**PDF of SLR** 

#### **Posterior informed (blue):** centered on positive tail of prior – not consistent with observations.

- Could be due to "ad hoc"  $\beta$  used as mean field (spin-up over 100 years).
- May be that emulator was been built with a (non-physical) positive mass balance while calibration was done on present-day observations (consistent with ice losing mass).

Results illustrate that we have in place all steps of our UQ workflow; they are NOT yet actual uncertainty bounds for sea-level rise. **Next step:** repeat UQ procedure with better modes, surrogates and  $\overline{\beta}$ .



#### Outline

*Albany/FELIX* = new land-ice solver with *next-generation* capabilities.

• Overview:

- First Order (FO) Stokes model.
- *Albany/FELIX* First-Order (FO) Stokes diagnostic solver.
- **CISM-Albany** and **MPAS-Albany** codes for prognostic simulations of the ice sheet evolution.
- Uncertainty Quantification (UQ):
  - Deterministic inversion.
  - Bayesian calibration.
  - Forward propagation of uncertainty.
- Performance portability.
- **Summary** and ongoing work.
- Questions?







# 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.).

- Kokkos: Trilinos library that provides performance portability across diverse devises with different memory models.
  - A *programming model* as much as a software library.
  - Provides automatic access to OpenMP, CUDA, Pthreads, etc.
  - Templated meta-programming: parallel\_for, parallel\_reduce (templated on an *execution space*).
  - Memory layout abstraction ("array of structs" vs. "struct of arrays", locality).

With *Kokkos*, you write an algorithm **once**, and just change a template parameter to get the optimal data layout for your hardware.







# 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.).

- Kokkos: Trilinos library that provides performance portability across diverse devises with different memory models.
  - A *programming model* as much as a software library.
  - Provides automatic access to OpenMP, CUDA, Pthreads, etc.
  - Templated meta-programming: parallel\_for, parallel\_reduce (templated on an *execution space*).
  - Memory layout abstraction ("array of structs" vs. "struct of arrays", locality).

With *Kokkos*, you write an algorithm **once**, and just change a template parameter to get the optimal data layout for your hardware.

• Finite element assembly in *Albany* has recently been rewritten using *Kokkos* functors.





# Kokkos-ification of Finite Element Assembly

```
typedef Kokkos::OpenMP ExecutionSpace;
//typedef Kokkos::CUDA ExecutionSpace;
//typedef Kokkos::Serial ExecutionSpace;
template<typename ScalarT>
vectorGrad<ScalarT>::vectorGrad()
Kokkos::View<ScalarT****, ExecutionSpace> vecGrad("vecGrad", numCells, numOP, numVec, numDim);
template<typename ScalarT>
void vectorGrad<ScalarT>::evaluateFields()
 Kokkos::parallel for<ExecutionSpace> (numCells, *this);
                                                                  ExecutionSpace parameter
template<typename ScalarT>
                                                                 tailors code for device (e.g.,
KOKKOS INLINE FUNCTION
void vectorGrad<ScalarT>:: operator() (const int cell) const
                                                                    OpenMP, CUDA, etc.)
ł
 for (int cell = 0; cell < numCells; cell++)</pre>
 for (int qp = 0; qp < numQP; qp++) {
    for (int dim = 0; dim < numVec; dim++) {</pre>
      for (int i = 0; i < numDim; i++) {
        for (int nd = 0; nd < numNode; nd++) {
          vecGrad(cell, qp, dim, i) += val(cell, nd, dim) * basisGrad(nd, qp, i);
```



# **Performance-Portability via** Kokkos: 20km GIS Problem

Sandia

National Laboratories



### Performance-Portability via *Kokkos*: Weak Scalability for GIS on *Titan*



#### Outline

*Albany/FELIX* = new land-ice solver with *next-generation* capabilities.

• Overview:

- First Order (FO) Stokes model.
- Albany/FELIX First-Order (FO) Stokes diagnostic solver.
- **CISM-Albany** and **MPAS-Albany** codes for prognostic simulations of the ice sheet evolution.
- Uncertainty Quantification (UQ):
  - Deterministic inversion.
  - Bayesian calibration.
  - Forward propagation of uncertainty.
- Performance portability.
- **Summary** and ongoing work.
- Questions?





# Summary and Ongoing Work

Summary: this talk described...

44

- The development of a finite element land ice solver known as *Albany/FELIX* written using the *Trilinos* libraries.
- Coupling of *Albany/FELIX* to the *CISM* and *MPAS LI* codes for transient simulations of ice sheet evolution.
- Advanced, next generation capabilities (UQ, performance portability) of Albany//FELIX.



#### **Ongoing/future work:**

- Science runs using CISM-Albany and MPAS-Albany.
- Deploy UQ workflow with better basis than KLE (e.g., Hessian eigenvectors).
- Continued porting of code to new architecture supercomputers (GPUs on Titan, Summit, Cori Phase I).
- Delivering code to climate community and coupling to earth system models.



#### Funding/Acknowledgements



45

**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.

Thank you! Questions?





nnal

ratories

[1] M.A. Heroux *et al.* "An overview of the Trilinos project." *ACM Trans. Math. Softw.* **31**(3) (2005).

[2] A.G. Salinger *et al.* "Albany: Using Agile Components to Develop a Flexible, Generic Multiphysics Analysis Code", *Comput. Sci. Disc.* (submitted, 2015).

46

[3] **I. Tezaur**, M. Perego, A. Salinger, R. Tuminaro, S. Price. "*Albany/FELIX*: A Parallel, Scalable and Robust Finite Element Higher-Order Stokes Ice Sheet Solver Built for Advanced Analysis", *Geosci. Model Develop.* 8 (2015) 1-24.

[4] **I. Tezaur**, R. Tuminaro, M. Perego, A. Salinger, S. Price. "On the scalability of the *Albany/FELIX* first-order Stokes approximation ice sheet solver for large-scale simulations of the Greenland and Antarctic ice sheets", *MSESM/ICCS15*, Reykjavik, Iceland (June 2014).

[5] R.S. Tuminaro, **I. Tezaur**, M. Perego, A.G. Salinger. "A Hybrid Operator Dependent Multi-Grid/Algebraic Multi-Grid Approach: Application to Ice Sheet Modeling", *SIAM J. Sci. Comput.* (in prep).

[6] R. Tuminaro. "ML's SemiCoarsening Feature, Addition to ML 5.0 Smoothed Aggregation User's Guide", Sandia National Laboratories Report, SAND2006-2649, Sandia National Laboratories, Albuquerque, NM, 2014.

# References (cont'd)

[7] S. Shannon, *et al.* "Enhanced basal lubrication and the contribution of the Greenland ice sheet to future sea-level rise", *P. Natl. Acad. Sci.*, 110 (2013) 14156-14161.

[8] P. Fretwell, *et al.* "BEDMAP2: Improved ice bed, surface, and thickness datasets for Antarctica", *The Cryosphere* 7(1) (2013) 375-393.

47

[9] F. Pattyn. "Antarctic subglacial conditions inferred from a hybrid ice sheet/ice stream model", *Earth and Planetary Science Letters 295 (2010).* 

[10] M. Perego, S. Price, G. Stadler. "Optimal Initial Conditions for Coupling Ice Sheet Models to Earth System Models", *J. Geophys. Res.* 119 (2014) 1894-1917.



# Appendix: Preliminary Results for GIS Bayesian Inference

• 5 KLE modes capture 95% of covariance energy  $\rightarrow$  parallel C++/Trilinos code (Anasazi).



• Mismatch = sum of squares of surface velocity discrepancy  $\rightarrow$  **Albany.** 

- Polynomial chaos expansion (PCE) was formed for the mismatch over  $\xi_k$  using uniform prior distributions and isotropic sparse grid level = 3  $\rightarrow$  **DAKOTA**.
- Markov Chain Monte Carlo (MCMC) was performed on the PCE with 100K samples → QUESO.
- <sup>16/18</sup> Collaborators: M. Perego, J. Jakeman, M. Eldred, L. Swiler (SNL)









### Appendix: Preliminary Results for GIS Bayesian Inference (cont'd)

Posterior distributions for the 5 KLF coefficients:



49





Coefficient 3





**MAP** solution:

Inference of KLE random field:







Sandia

National

Laboratories

17/18