

Adjoint-based deterministic Inversion for Ice Sheets

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joint work with:

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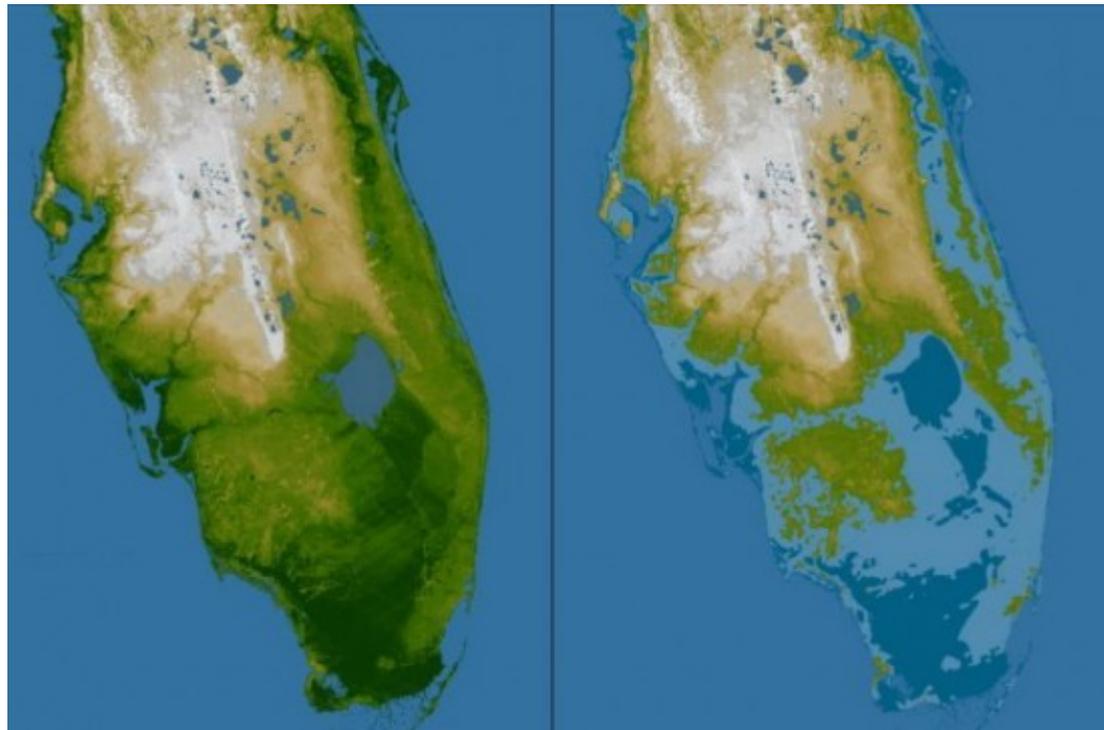
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Motivations

- Glaciers and ice sheets influence the global climate, and vice-versa
- Melting of land ice determines the **sea level rise**
 - melting of the Greenland ice sheet: 7 m
 - melting of the Antarctic ice sheet: 61 m



South Florida projection for a sea levels rise of 5m (dark blue) and 10m (light blue)

Ice Sheet Modeling

Main components of an ice model:

- **Ice flow equations** (momentum and mass balance)

$$\begin{cases} \rho \left(\frac{\partial \mathbf{u}}{\partial t} + \mathbf{u} \cdot \nabla \mathbf{u} \right) - \nabla \cdot \sigma = \rho \mathbf{g} \\ \nabla \cdot \mathbf{u} = 0 \end{cases}$$



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$$\begin{cases} -\nabla \cdot \sigma = \rho \mathbf{g} \\ \nabla \cdot \mathbf{u} = 0 \end{cases}$$

with:

$$\sigma = 2\mu \mathbf{D} - \Phi I, \quad \mathbf{D}_{ij}(\mathbf{u}) = \frac{1}{2} \left(\frac{\partial u_i}{\partial x_j} + \frac{\partial u_j}{\partial x_i} \right)$$

Non linear viscosity:

$$\mu = \frac{1}{2} \alpha(T) |\mathbf{D}(\mathbf{u})|^{(p-2)}, \quad p \in (1, 2] \quad (\text{typically } p \simeq \frac{4}{3})$$



Viscosity is singular when ice is not deforming



Ice Sheet Modeling

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Ice Sheet Modeling

Main components of an ice model:

- **Ice flow equations** (momentum and mass balance)

$$\begin{cases} -\nabla \cdot \sigma = \rho \mathbf{g} \\ \nabla \cdot \mathbf{u} = 0 \end{cases}$$

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

$$\frac{\partial H}{\partial t} = H_{flux} - \nabla \cdot \int_z \mathbf{u} dz$$

- **Temperature equation**

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

- **Coupling with other climate components** (e.g. ocean, atmosphere)



Stokes Approximations

“Reference” model: **STOKES**¹

$O(\delta^2)$ **FO**, Blatter-Pattyn first order model² (3D PDE, in horizontal velocities)

$O(\delta)$ Zeroth order, depth integrated models:
SIA, Shallow Ice Approximation (slow sliding regimes) ,
SSA Shallow Shelf Approximation (2D PDE) (fast sliding regimes)

$\simeq O(\delta^2)$ Higher order, depth integrated (2D) models: **L1L2**³, (L1L1)...

$\delta :=$ ratio between ice thickness and ice horizontal extension

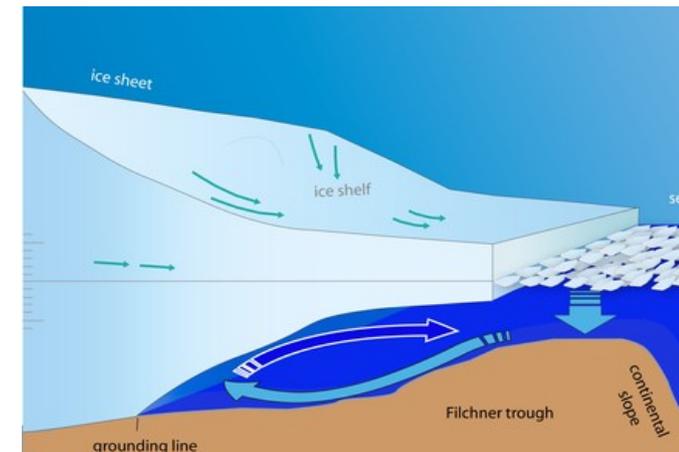
¹Gagliardini and Zwinger, 2008. *The Cryosphere*.

²Dukowicz, Price and Lipscomb, 2010. *J. Glaciol.*

³Schoof and Hindmarsh, 2010. *Q. J. Mech. Appl. Math.*

(Numerical) Modeling Issues

- Computationally challenging, due to complexity of models, of geometries and large domains
 - design of linear/nonlinear solvers, preconditioners, etc.
 - mesh adaptivity especially close to the grounding line.
- Boundary conditions / coupling (e.g. with ocean)
 - Floating/calving
 - Basal friction at the bedrock,
 - Subglacial hydrology,
 - Heat exchange / phase change.
- Initialization / parameter estimation.
- Uncertainty quantification.



Inverse Problem

Estimation of ice-sheet initial state*

(w/ G. Stadler, UT, and S. Price, LANL)

Problem: what is the initial thermo-mechanical state of the ice sheet?

Available data/measurements:

- *ice extension and surface topography*
- *surface velocity*
- *Surface Mass Balance (SMB: accumulation/melt rate)*
- *ice thickness H (very noisy)*

Fields to be estimated :

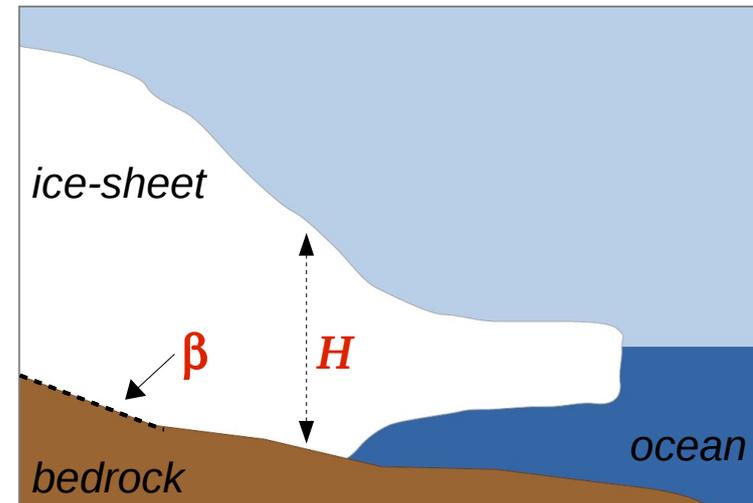
- ***ice thickness H***
- ***basal friction β***

Additional information:

- *ice fulfills **nonlinear Stokes equation***
- *ice is almost **at thermo-mechanical equilibrium***

Assumption (for now):

- *given **temperature field***



Inverse Problem

Estimation of ice-sheet initial state

Steady State equations and basal sliding conditions

How to prescribe ice-sheet mechanical equilibrium:

$$\frac{\partial H}{\partial t} = -\text{div}(\mathbf{U}H) + \tau_s, \quad \mathbf{U} = \frac{1}{H} \int_z \mathbf{u} dz.$$

divergence flux (pointing to $\mathbf{U}H$)
Surface Mass Balance (pointing to τ_s)

At equilibrium: $\text{div}(\mathbf{U}H) = \tau_s$

Boundary condition at ice-bedrock interface:

$$(\sigma \mathbf{n} + \beta \mathbf{u})_{\parallel} = \mathbf{0} \quad \text{on} \quad \Gamma_{\beta}$$

Bibliography*:

- Arthern, Gudmundsson, J. Glaciology. 2010*
- Price, Payne, Howat and Smith, PNAS 2011*
- Brinkerhoff et al., Annals of Glaciology, 2011*
- Morlighem et al. Geophysical Research Letters, 2013*
- Pollard DeConto, TCD 2012*
- Petra, Zhu, Stadler, Hughes, Ghattas, J. Glaciology, 2012.*
- Goldberg and Heimbach, The Cryosphere 2013.*
- Michel et al., Computers & Geosciences, 2014.*



Inverse Problem

Estimation of ice-sheet initial state

PDE-constraint optimization problem: cost functional

Problem: find initial conditions such that the ice is almost at thermo-mechanical equilibrium given the geometry and the SMB, and matches available observations.

Optimization Problem:

find β and H that minimizes the functional \mathcal{J}

$$\begin{aligned}\mathcal{J}(\beta, H) &= \int_{\Sigma} \frac{1}{\sigma_u^2} |\mathbf{u} - \mathbf{u}^{obs}|^2 ds \\ &+ \int_{\Sigma} \frac{1}{\sigma_{\tau}^2} |\operatorname{div}(\mathbf{U}H) - \tau_s|^2 ds \\ &+ \int_{\Sigma} \frac{1}{\sigma_H^2} |H - H^{obs}|^2 ds \\ &+ \mathcal{R}(\beta, H)\end{aligned}$$

surface velocity mismatch } } *Common*
SMB mismatch } } *Proposed*
thickness mismatch }

regularization terms.

subject to ice sheet model equations
(FO or Stokes)

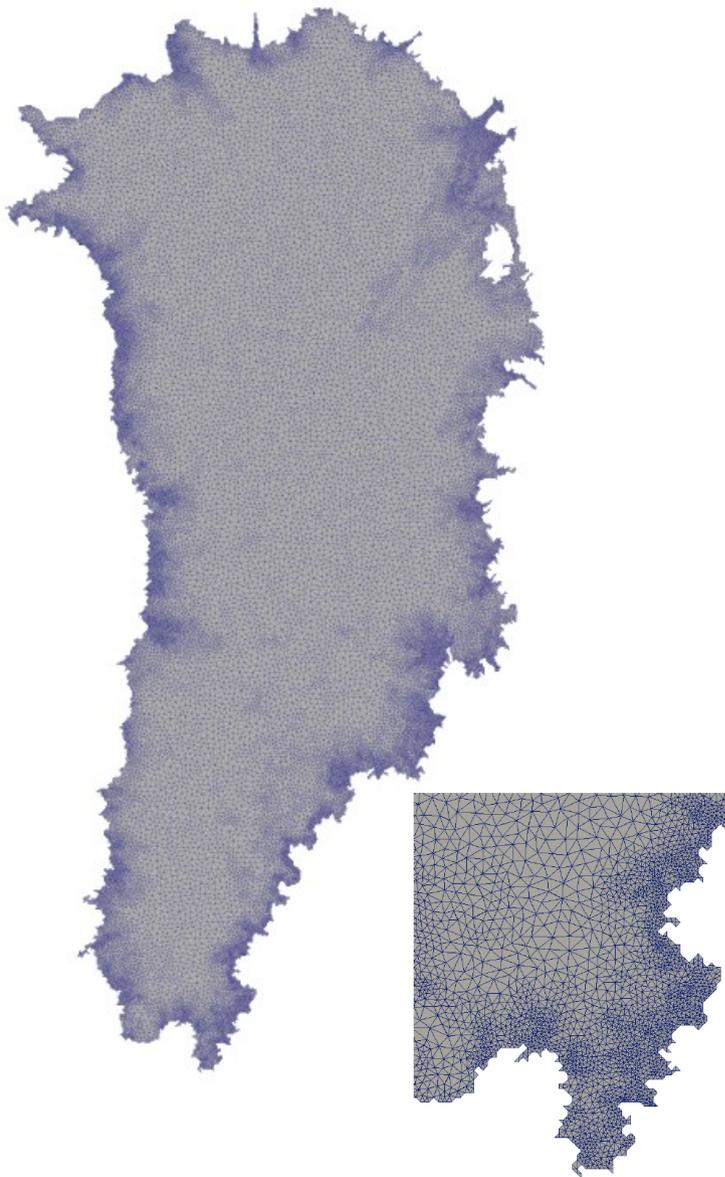
\mathbf{U} : computed depth averaged velocity
 H : ice thickness
 β : basal sliding friction coefficient
 τ_s : SMB
 $\mathcal{R}(\beta)$ regularization term

Inverse Problem

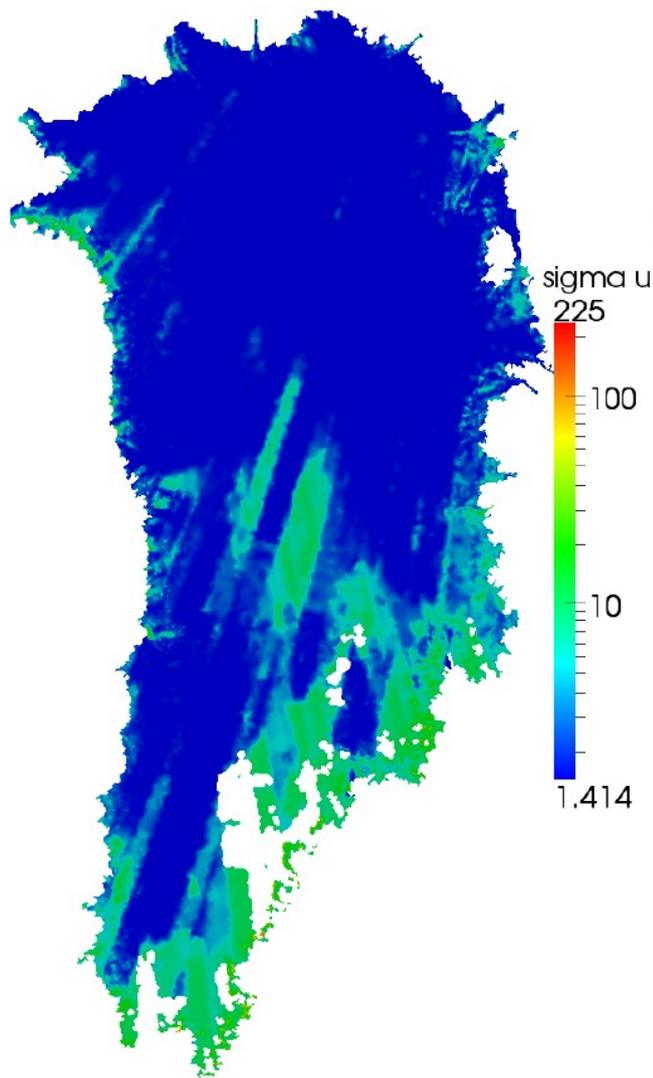
Estimation of ice-sheet initial state of Greenland ice sheet

Grid and RMS of velocity and errors associated with velocity and thickness observations

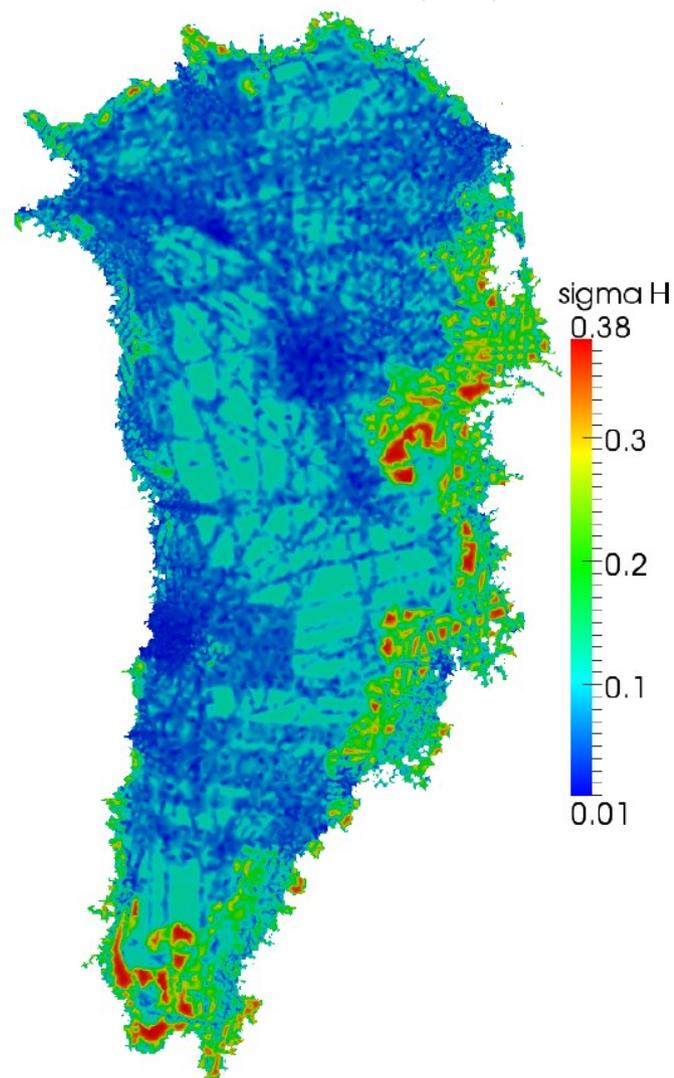
Grid



Velocity RMS (m/yr)



Thickness RMS (km)



Inverse Problem

Estimation of ice-sheet initial state of Greenland ice sheet

Inversion results: surface velocities

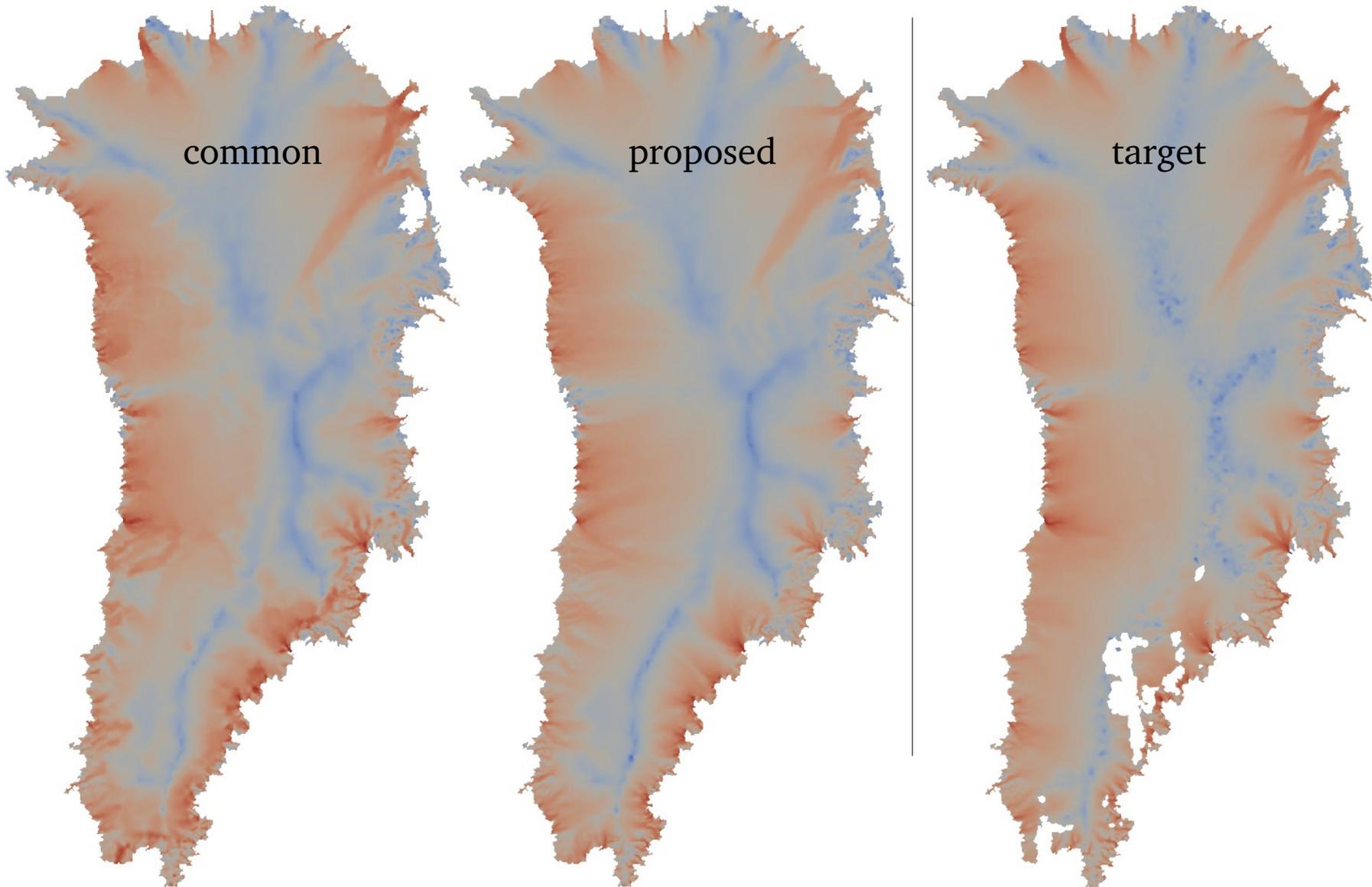
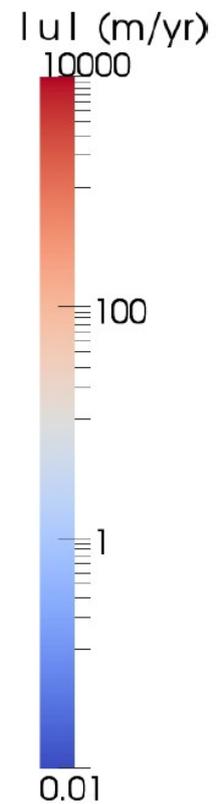
computed surface velocity

observed surface velocity

common

proposed

target

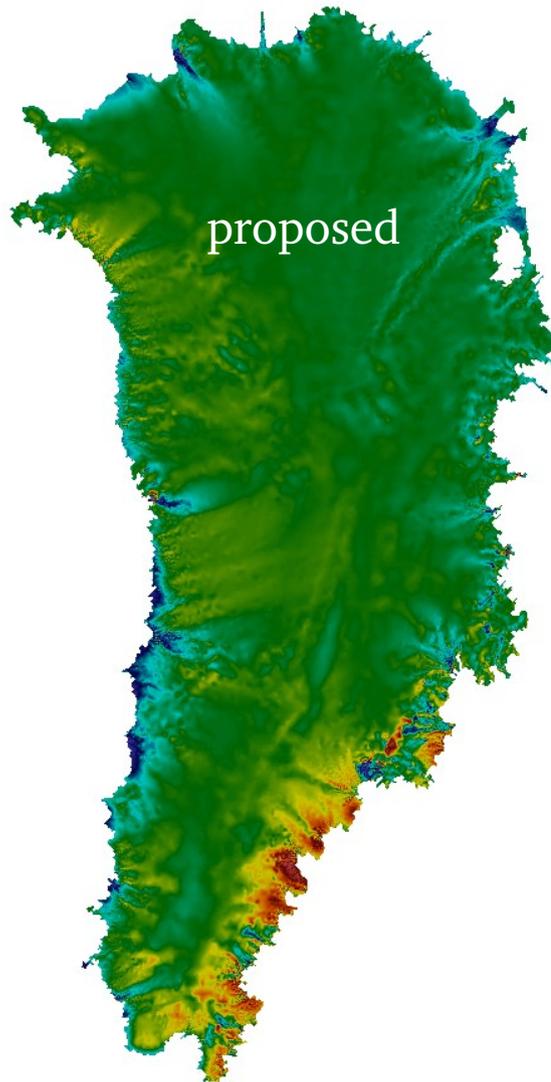
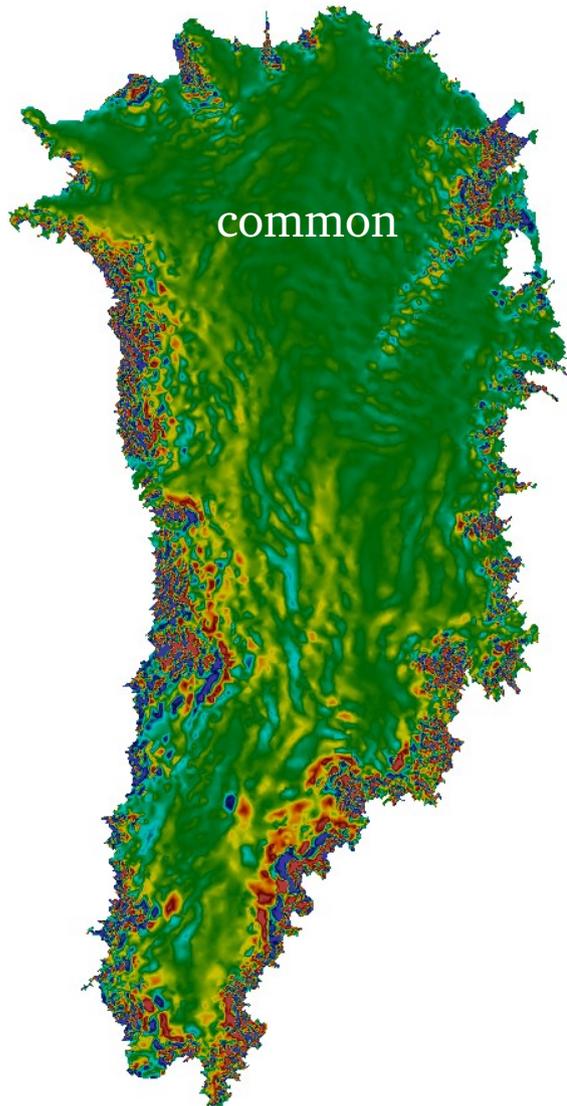


Inverse Problem

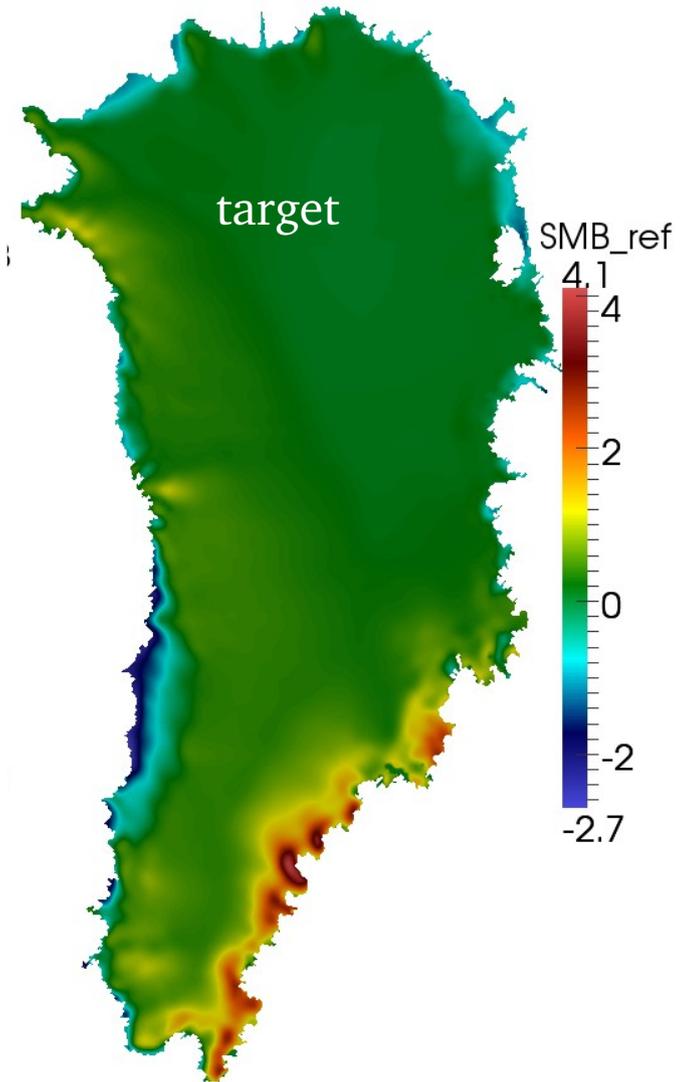
Estimation of ice-sheet initial state of Greenland ice sheet

Inversion results: surface mass balance (SMB)

SMB needed for equilibrium



SMB from climate model



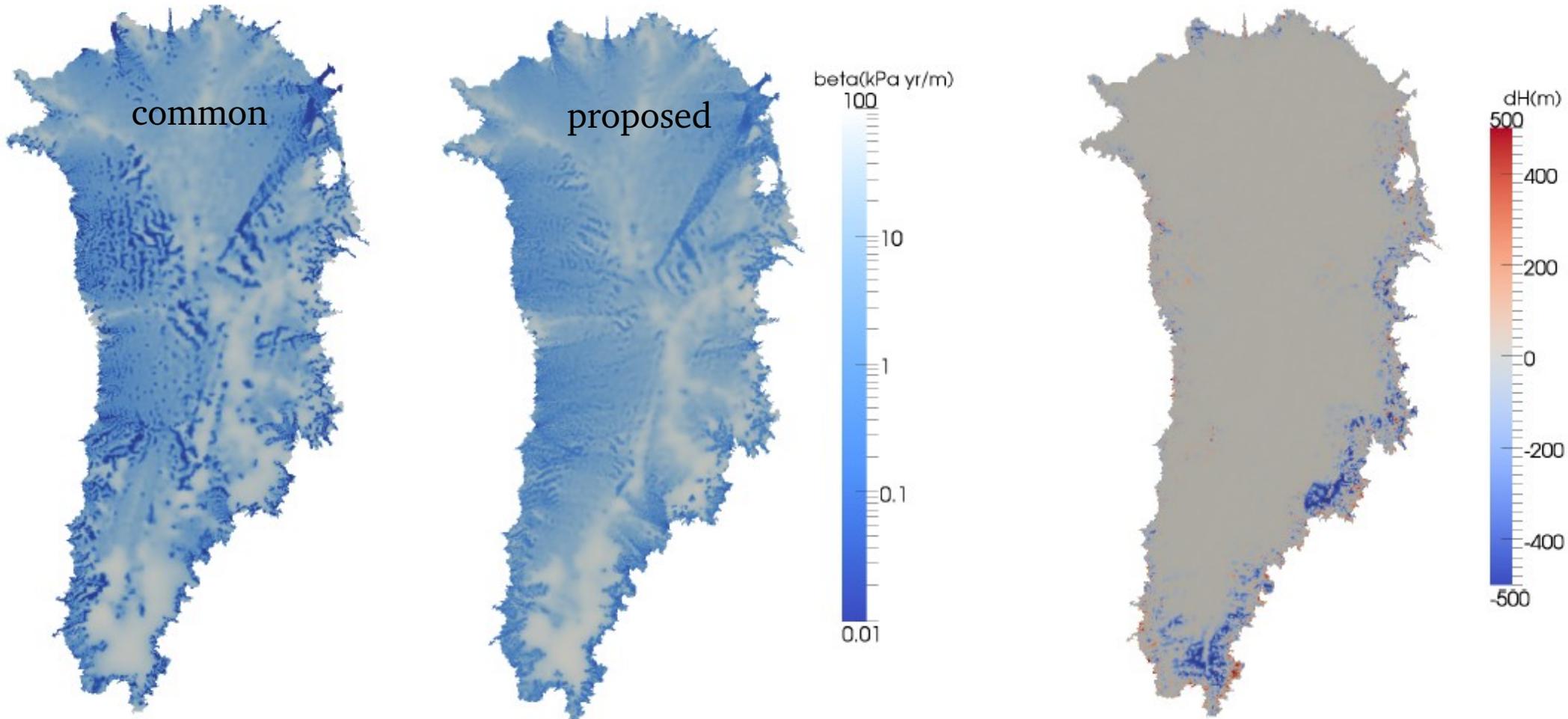
Inverse Problem

Estimation of ice-sheet initial state of Greenland ice sheet

Estimated beta and change in topography.

recovered basal friction

difference between recovered
and observed thickness





Inverse Problem

Estimation of ice-sheet initial state

Algorithm and Software tools used

| Algorithm | Software Tools |
|--------------------------------------|-----------------------|
| Basal non-uniform triangular mesh | <i>Triangle</i> |
| Linear Finite Elements on tetrahedra | <i>LifeV</i> |
| Quasi-Newton optimization (L-BFGS) | Rol |
| Nonlinear solver (Newton method) | NOX |
| Krylov Linear Solvers | AztecOO/IfPack |



Details:
Regularization terms: Tikhonov.
L-BFGS initialized with Hessian of the regularization terms.

$$\left(\frac{1}{2} \beta^T L \beta \quad \rightarrow \quad L \right)$$

Inverse Problem

Estimation of ice-sheet initial state

PDE-constraint optimization problem: gradient computation

Find (β) that minimize $\mathcal{J}(\beta, \mathbf{u})$
subject to $\mathcal{F}(\mathbf{u}, \beta) = 0.$

← flow model

How to compute **total derivatives** of the functional w.r.t. the parameters?

Solve State System $\mathcal{F}(\mathbf{u}, \beta) = 0$

Solve Adjoint System $\langle \mathcal{F}_{\mathbf{u}}^*(\boldsymbol{\lambda}), \boldsymbol{\delta}_{\mathbf{u}} \rangle = \mathcal{J}_{\mathbf{u}}(\boldsymbol{\delta}_{\mathbf{u}}), \quad \forall \boldsymbol{\delta}_{\mathbf{u}}$

Total derivative $\mathcal{G}(\delta_{\beta}) = \mathcal{J}_{\beta}(\delta_{\beta}) - \langle \boldsymbol{\lambda}, \mathcal{F}_{\beta}(\delta_{\beta}) \rangle$

Derivative w.r.t. β

$$\mathcal{G}_1(\delta_{\beta}) = \alpha_{\beta} \int_{\Sigma} \nabla \beta \cdot \nabla \delta_{\beta} \, ds - \int_{\Sigma} \delta_{\beta} \mathbf{u} \cdot \boldsymbol{\lambda} \, ds$$



Porting the inversion to Albany-FELIX

(w/ E. Phipps, A. Salinger, D. Ridzal and D. Kouri)

Why?

- to exploit Automatic Differentiation for computing derivatives
- to exploit Albany/Trilinos ecosystem (e.g. for UQ capabilities using Dakota)
- to extend Albany adjoint/inversion capabilities,
- to use in-house software (better maintainability)

Albany Development:

- implement **distributed parameters**, i.e. fields defined on the mesh or on parts of it.
- implement routines for computing **derivatives** of *residual* and *responses* w.r.t. the **distributed parameters**.

Trilinos Development:

- couple Piro to ROL using Thyra implementation of ROL::Vector and ROL::Objective.
ROL needs reduced gradient and objective functional.

$$\mathcal{G} = \mathcal{J}_\beta - \mathcal{F}_\beta^T \lambda$$

Matrix-free
matrix vector product

\mathcal{G} : reduced gradient

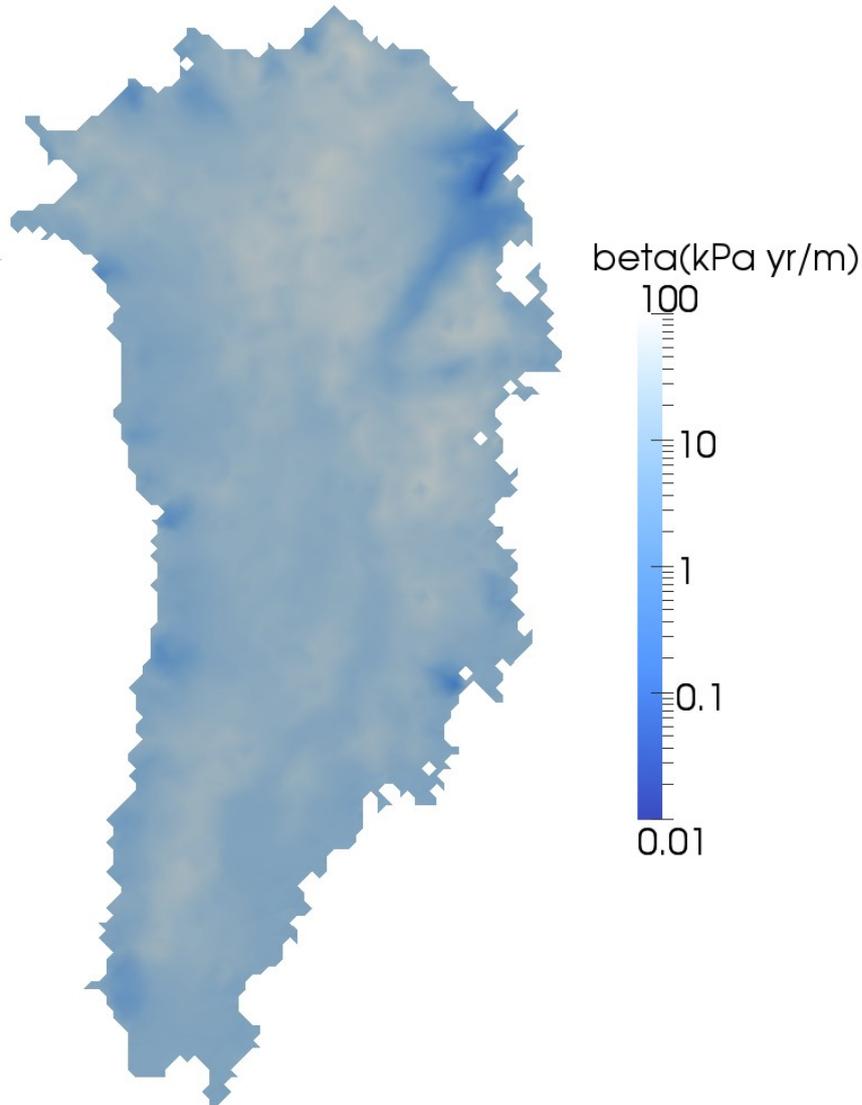
\mathcal{J} : response or objective function

\mathcal{F} : residual

β : (distributed) parameter

Preliminary result using Albany-Piro-ROL

recovered basal friction



Objective functional:

$$\mathcal{J}(\mathbf{u}(\beta), \beta) = \int_{\Sigma} \frac{1}{\sigma_u^2} |\mathbf{u} - \mathbf{u}^{obs}|^2 ds + \alpha \int_{\Sigma} |\nabla \beta|^2 ds.$$

ROL algorithm:

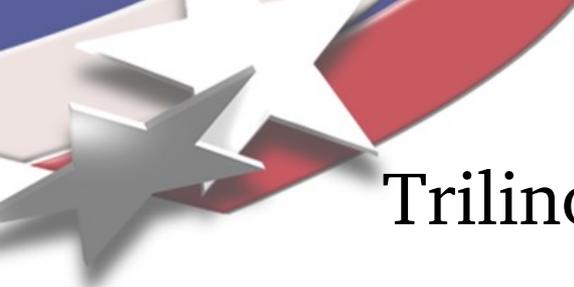
- *Limited-Memory BFGS*
 - *Backtrack line-search*
- Inverted 2000 parameters.*

TODOs:

- clean/test Piro-ROL interface
- add bound-constraints

- implement Hessian computation in order to use Newton methods and for UQ

- invert for shape parameters (H)



Trilinos packages used in this calculation:

Nonlinear analysis

- Piro
- ROL
- LOCA
- NOX

Vectors / tools

- Teuchos
- Epetra
- EpetraExt
- Thyra

Physics / discretization

- Phalanx
- Sacado
- Intrepid
- Shards

Linear solvers

- Stratimikos
- Belos/AztecOO
- Ifpack

Mesh

- Zoltan
- STK::Mesh
- STK::IO
- Seacas::Exodus
- Seacas::IOSS

(thanks Andy)