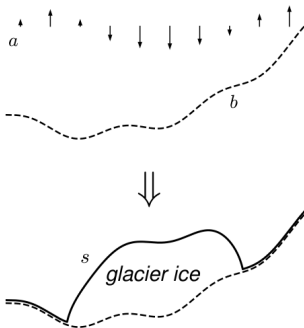


IMPLICIT MULTIGRID METHODS FOR GLACIER GEOMETRY

ED BUELER

SIAM AN22



UNIVERSITY
of ALASKA
Many Traditions One Alaska

1. model for evolving glacier surfaces
2. implicit-step free-boundary problem (NCP or VI)
3. computational performance model
4. geometric multigrid with distinctive features
 - ▶ constraint decomposition
 - ▶ FAS-type nonlinear coarse corrections
5. minimal results, smoother difficulties

THE GLACIER GEOMETRY PROBLEM

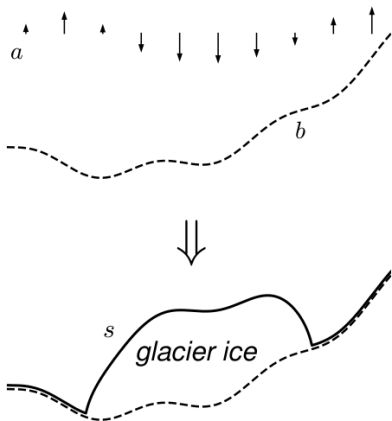
■ inputs:

- ▶ a = climatic mass balance (snowfall or melt&runoff)
- ▶ b = bed elevation
- ▶ other inputs ignored for simplicity: surface temperature, bed composition, geothermal, ...

■ goal: **construct fast numerical methods for this map!**

(climate and topography) \rightarrow (glacier geometry)

$a, b \mapsto s$



GLACIERS FLOW

- glacier geometry would be easy if the snow just piled-up!
- but glaciers flow
 - ▶ very viscous, non-Newtonian fluid driven by gravity
 - ▶ flow is mostly downhill, along $-\nabla_x s$
 - ▶ ice flows into areas where there is melt



- $\mathbf{x} = (x, y)$ denotes horizontal coordinates
 - ▶ \mathbf{x} is map-plane
- data given on 2D domain Ω :
 - ▶ climate (time-dependent): $a(t, \mathbf{x})$
 - ▶ bed elevation (stationary): $b(\mathbf{x})$
- we seek surface elevation $s(t, \mathbf{x})$
 - ▶ obviously glaciers are on top: $s(t, \mathbf{x}) \geq b(\mathbf{x})$
- also seek velocity $\mathbf{u}(t, \mathbf{x}, z)$ within 3D ice
 - ▶ $\Lambda_s := \{b(\mathbf{x}) < z < s(t, \mathbf{x})\}$

SURFACE KINEMATICAL EQUATION (SKE)

- mass conservation at glacier surface?
- this is the **surface kinematical equation**:

$$\frac{\partial s}{\partial t} - \mathbf{u}|_s \cdot \mathbf{n}_s - a = 0$$

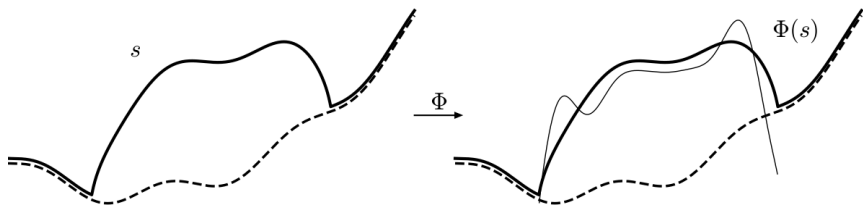
- ▶ $\mathbf{n}_s = \langle -\nabla_{\mathbf{x}}s, 1 \rangle$ is upward normal to surface
- ▶ $\mathbf{u}|_s = \mathbf{u}(t, \mathbf{x}, s(t, \mathbf{x}))$ is 3D velocity at surface

ABSTRACT THE FLOW

- we will regard flow as a Stokes problem
 - ▶ or a shallow approximation thereof
- problem becomes clearer if we **abstract the flow** as a map:

$$\left(\begin{array}{c} \text{surface} \\ \text{elevation} \end{array} \right) \rightarrow \left(\begin{array}{c} \text{ice motion} \\ \text{from dynamics} \\ \text{at surface} \end{array} \right)$$

$$\Phi : s \mapsto -\mathbf{u}|_s \cdot \mathbf{n}_s$$



HOW TO COMPUTE $\Phi(s)$?

- how to compute $\Phi(s) = -\mathbf{u}|_s \cdot \mathbf{n}_s$?

- **Stokes**: solve for \mathbf{u} :

$$\int_{\Lambda_s} 2\nu_\epsilon(D\mathbf{u})D\mathbf{u} : D\mathbf{v} - p\nabla \cdot \mathbf{v} - (\nabla \cdot \mathbf{u})q - \rho_i \mathbf{g} \cdot \mathbf{v} \, d\mathbf{x} = 0,$$

over all test \mathbf{v}, q , where $\Lambda_s = \{(\mathbf{x}, z) : b(\mathbf{x}) < z < s(t, \mathbf{x})\}$ is the 3D ice and

$$\nu_\epsilon(D\mathbf{u}) = \frac{1}{2} \Gamma (|\mathbf{D}\mathbf{u}|^2 + \epsilon D_0^2)^{(p-2)/2},$$

is effective viscosity and $p = \frac{1}{n} + 1$, then extract trace $\mathbf{u}|_s$, then extend $\Phi(s) = -\mathbf{u}|_s \cdot \mathbf{n}_s$ by 0 to Ω

- ▶ well-posed over $W_0^{1,p}(\Lambda_s)^3 \times L^q(\Lambda_s)$ (Jouvet & Rappaz 2011)
- ▶ near-optimal solver (Isaac et al 2015)

HOW TO COMPUTE $\Phi(\mathbf{s})$?

- how to compute $\Phi(\mathbf{s}) = -\mathbf{u}|_s \cdot \mathbf{n}_s$?
- shallow ice approximation (SIA): evaluate

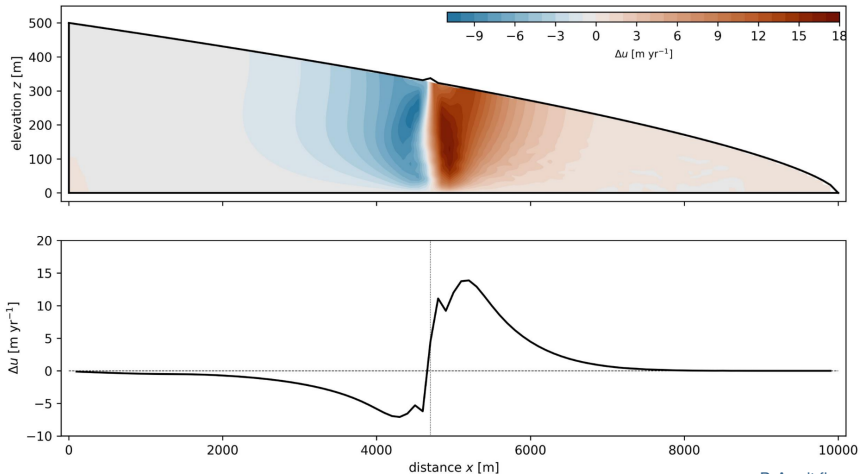
$$\Phi(\mathbf{s}) = -\frac{\gamma}{q}(\mathbf{s}-b)^q |\nabla_{\mathbf{x}} \mathbf{s}|^q - \nabla_{\mathbf{x}} \cdot \left(\frac{\gamma}{q+1} (\mathbf{s}-b)^{q+1} |\nabla_{\mathbf{x}} \mathbf{s}|^{q-2} \nabla_{\mathbf{x}} \mathbf{s} \right)$$

where $q = n + 1$ and $\gamma > 0$ is related to ice softness and density

- ▶ SIA = lubrication approximation
- ▶ $\Phi(\mathbf{s})$ is a nonlinear differential operator only because membrane (longitudinal) stresses are unbalanced

VELOCITY EXPRESSION OF S PERTURBATION

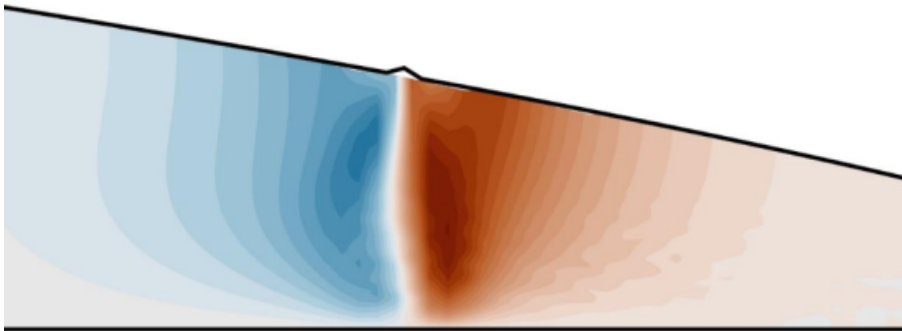
- derivative of $\Phi(s) = -\mathbf{u}|_s \cdot \mathbf{n}_s$?
- difference $\mathbf{u}_{(s+\psi)} - \mathbf{u}_{(s)}$ from surface perturbation ψ



P. Arndt figure

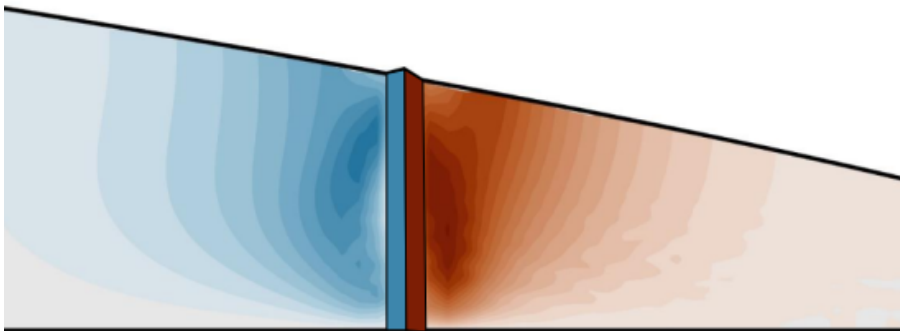
A CLOSER LOOK

- perturb s by 5 m bump, over 200 m
- resulting **Stokes velocity difference**
 - ▶ Stokes $\Phi(s)$ is not local
 - ▶ note longitudinal-stress range (Kamb & Echelmeyer 1986)



A CLOSER LOOK

- perturb s by 5 m bump, over 200 m
- resulting **Stokes velocity difference**
 - ▶ Stokes $\Phi(s)$ is not local
 - ▶ note longitudinal-stress range (Kamb & Echelmeyer 1986)
 - ▶ overlain **SIA difference**, localized under s perturbation!



- goal: determine glacier surface elevation s and **glaciated area**
- this is a free-boundary problem in the map-plane
- basic logic gives a **nonlinear complementarity*** problem (NCP) over all of Ω :

$$\begin{aligned} s - b &\geq 0 \\ \frac{\partial s}{\partial t} + \Phi(s) - a &\geq 0 \\ (s - b) \left(\frac{\partial s}{\partial t} + \Phi(s) - a \right) &\stackrel{*}{=} 0 \end{aligned}$$

- ▶ we extend $\mathbf{u}|_s$ by zero so $\Phi(s) = 0$ in ice-free areas
- ▶ glaciated area $\Omega_+(t) := \{\mathbf{x} : s(t, \mathbf{x}) > b(\mathbf{x})\} \dots$ from solution!

BACKWARD EULER STEP

- ice mostly flows downhill, so SKE is mostly diffusive!
 - ▶ straightforward in shallow ice approximation
 - ▶ not literally a diffusion with Stokes dynamics
- implicit stepping makes sense . . . try **backward Euler**:

$$\frac{s - s_0}{\Delta t} + \Phi(s) - a = 0$$

- ▶ s_0 is the previous surface elevation
 - ▶ s is new surface elevation
- each time step is a free boundary problem, an NCP:

$$\begin{aligned} s - b &\geq 0 \\ s + \Delta t \Phi(s) - (s_0 + \Delta t a) &\geq 0 \\ (s - b) \left(s + \Delta t \Phi(s) - (s_0 + \Delta t a) \right) &= 0 \end{aligned}$$

- convert to weak form for FE treatment: $\text{NCP} \implies \text{VI}$
- define **admissible surface elevations**:

$$\mathcal{K} = \{r : r \geq b\}$$

- ▶ the **constraint set**
- ▶ closed and convex subset of $\mathcal{V} = W^{1,p}(\Omega)$
- define nonlinear form and source linear functional:

$$N(s)[q] = \int_{\Omega} (s + \Delta t \Phi(s)) q \, d\mathbf{x}, \quad f[q] = \int_{\Omega} (s_0 + \Delta t a) q \, d\mathbf{x}$$

- **backward Euler step VI**: find $s \in \mathcal{K}$ so that

$$N(s)[r - s] \geq f[r - s]$$

for all $r \in \mathcal{K}$

DISCRETIZATION AND SOLVERS?

- VI = abstract, dynamics-agnostic, implicit time-step problem:

$$N(s)[r - s] \geq f[r - s]$$

- next steps: FE discretization, solvers
- can we actually solve it?
 - A. **yes** for the SIA (Bueler 2016)
 - A. efficiently, some day, I hope, for Stokes
- what do current models do?
 - A. explicit time stepping, usually forward Euler, **and that is a tragedy**

PISM
PARALLEL ICE SHEET MODEL



elmer
ICE



COMPUTATIONAL PARAMETERS

- for this story I need a performance model (Bueler, submitted)
- use only 3 parameters to describe simulations: typical values
 - ▶ climate-couplings per model year q 0.1 – 12
 - ▶ domain size (length) L 1000 km – 10 km
 - ▶ mesh spacing Δx 10 km – 10 m
- **stability restrictions for explicit schemes:**
 - ▶ optimistic (CFL): $\Delta t < O\left(\frac{\Delta x}{U}\right)$ weeks to hours?
 - ▶ pessimistic: $\Delta t < O\left(\frac{\Delta x^2}{D}\right)$ days to minutes?
- degrees of freedom m scale with 2D mesh spacing:

$$\Delta x = O\left(\frac{L}{\sqrt{m}}\right) \iff m = O\left(\frac{L^2}{\Delta x^2}\right)$$

ALGORITHMIC SCALING

■ algorithmic scaling of solvers:

- ▶ fixed-geometry velocity solution: $O(m^{1+\alpha})$ flops

$\alpha \approx 0$ (multigrid: Brown, Isaac, Tuminaro), 1 (sparse direct)

- ▶ implicit (velocity&geometry) solution: $O(m^{1+\gamma})$ flops

$\gamma \approx$ unknown (Stokes), 0.8 (SIA: Bueler 2016)

■ giving a model for **asymptotic simulation cost**:

time-stepping

flops per model year

explicit (optimistic) $O\left(\frac{UL^{2+2\alpha}}{\Delta x^{3+2\alpha}}\right) = O\left(\frac{Um^{1.5+\alpha}}{L}\right)$

explicit (pessimistic) $O\left(\frac{DL^{2+2\alpha}}{\Delta x^{4+2\alpha}}\right) = O\left(\frac{Dm^{2+\alpha}}{L^2}\right)$

implicit $O\left(\frac{qL^{2+2\gamma}}{\Delta x^{2+2\gamma}}\right) = O\left(qm^{1+\gamma}\right)$

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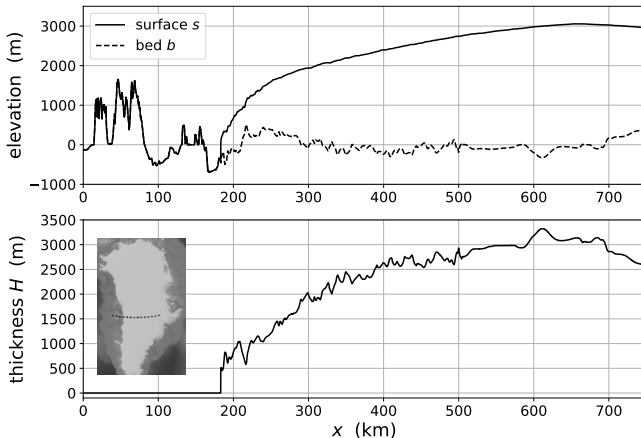
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← tragedy

$$\text{implicit} \quad O\left(\frac{qL^{2+2\gamma}}{\Delta x^{2+2\gamma}}\right) = O\left(qm^{1+\gamma}\right)$$

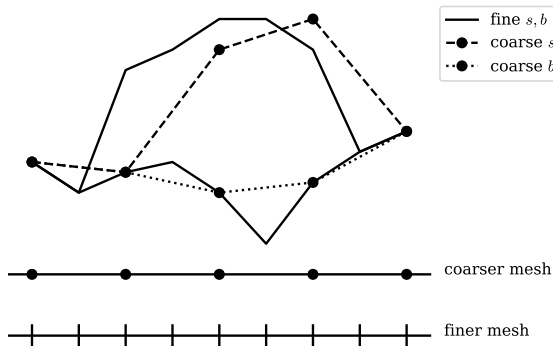
GEOMETRIC MULTIGRID FOR IMPLICIT STEPS?

- goal: $\gamma \approx 0$ in $O(m^{1+\gamma})$ cost of implicit solve
- can GMG be used as a solver?
 - ▶ need exploitable solution property: surface smoothness



GEOMETRIC MULTIGRID FOR IMPLICIT STEPS?

- solving a free-boundary (obstacle) problem
 - ▶ constraint $s \geq b$ (obstacle is rough bed elevation data!)
- issues:
 1. fine mesh bed data not present on coarse mesh
 - coarse iterates not necessarily admissible on fine
 2. free boundary location depends on mesh



- next: outline a GMG for the glacier geometry problem
 - ▶ simultaneous solution for surface elevation and velocity

- distinctive flavors:
 1. subspace decomposition viewpoint
 2. ordinary residual does not converge to zero
 3. each coarse level has 2 constraint sets
 - defect constraint
 - monotone restriction
 - multilevel constraint decomposition (Tai, 2003)
 4. FAS-type nonlinear coarse corrections
 5. smoother?
 - Stokes residual is expensive and non-local!

SUBSPACE DECOMPOSITION VIEWPOINT

- hat functions span FE space on level j :

$$\mathcal{V}^j = \text{span}\{\psi_1^j(\mathbf{x}), \dots, \psi_{m_j}^j(\mathbf{x})\}$$

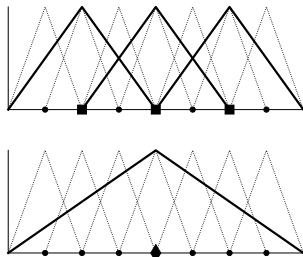
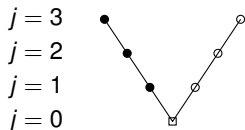
- **multilevel subspace decomposition:**

$$\mathcal{V}^h = \mathcal{V}^0 + \mathcal{V}^1 + \dots + \mathcal{V}^J$$

- hats are combinations of finer hats:

$$\psi_p^{j-1}(\mathbf{x}) = \sum_{q=1}^{m_j} c_{pq} \psi_q^j(\mathbf{x})$$

- ▶ $c_{pq} = \psi_p^{j-1}(\mathbf{x}_q^j)$
- ▶ canonical prolongation P
- ▶ canonical (dual) restriction R
- ▶ see Xu (1992)



- consider abstract NCP

$$\begin{aligned} s - b &\geq 0 \\ F(s) &\geq 0 \\ (s - b)F(s) &= 0 \end{aligned}$$

- the solution **does not satisfy** $F(s) = 0$ everywhere
 - ▶ generally $F(s) > 0$ where $s = b$
- how to tell if an iterate $w \approx s$ is converged?
- A. need to monitor **NCP residual** $\hat{F}(w)$:

$$\hat{F}(w)[\psi_p^J] = \begin{cases} F(w)[\psi_p^J], & w(\mathbf{x}_p^J) > b(\mathbf{x}_p^J) \\ \min\{F(w)[\psi_p^J], 0\}, & w(\mathbf{x}_p^J) = b(\mathbf{x}_p^J) \end{cases}$$

$$\|\hat{F}(w)\|_{(\mathcal{V}^J)'} < \text{tol} \quad \implies \quad w \text{ solves NCP}$$

■ suppose:

- ▶ b^j is fine-level bed elevation
- ▶ w^j is admissible fine-level iterate ($w^j \geq b^j$)

■ define **defect constraint** and **defect constraint set**:

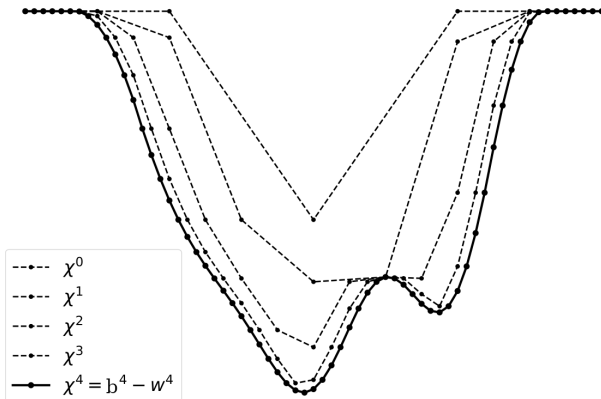
$$\chi^j = b^j - w^j$$

$$\mathcal{D}^j = \{v \geq \chi^j\} \subset \mathcal{V}^j$$

- ▶ notice $\chi^j \leq 0$
- ▶ $w^j + z^j$ is admissible if and only if $z^j \in \mathcal{D}^j$

DECOMPOSE THE DEFECT CONSTRAINT

- Tai (2003): decompose the fine-level defect constraint onto coarser meshes using monotone restriction



DECOMPOSE THE DEFECT CONSTRAINT

- monotone restriction operator $R^\oplus : \mathcal{V}^j \rightarrow \mathcal{V}^{j-1}$:

$$R^\oplus z = \sum_{p=1}^{m_{j-1}} \max\{z_p : \psi_p^{j-1}(x_p^j) > 0\} \psi_p^{j-1}$$

▶ observe $R^\oplus z \geq z$

- for $j = 1, \dots, J$ let

$$\chi^{j-1} = R^\oplus \chi^j$$

▶ also define $\chi^{-1} = 0$

- gaps between defect constraints: $\phi^j = \chi^j - \chi^{j-1}$

- telescoping-sum:

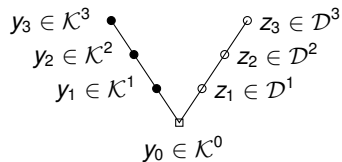
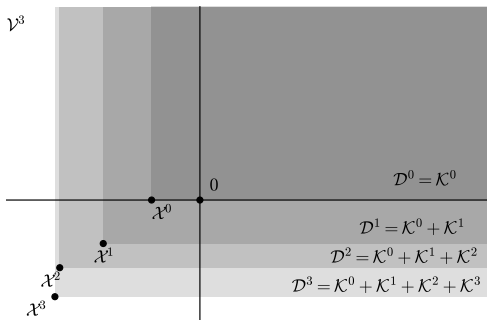
$$\sum_{j=0}^J \phi^j = \chi^0 + (\chi^1 - \chi^0) + (\chi^2 - \chi^1) + \dots + (\chi^J - \chi^{J-1}) = \chi^J$$

MULTILEVEL CONSTRAINT DECOMPOSITION (MCD)

- let

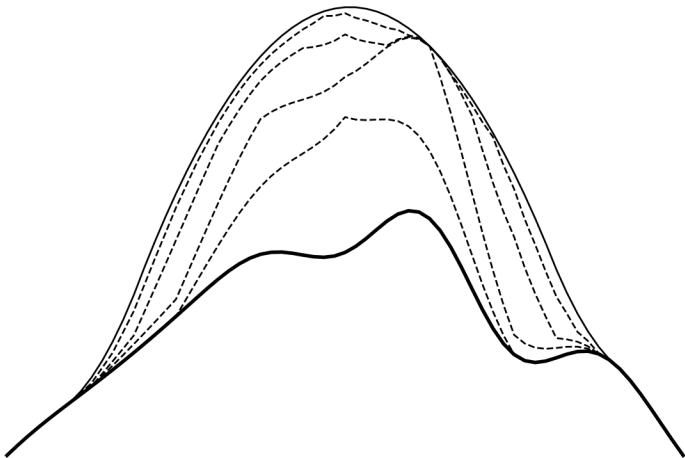
$$\mathcal{D}^j = \{v \geq \chi^j\}, \quad \mathcal{K}^j = \{v \geq \phi^j\}$$

- Tai (2003): decomposition of the fine-level defect constraint set \mathcal{D}^J by **cones** $\mathcal{D}^J = \mathcal{K}^0 + \dots + \mathcal{K}^J$ from the inside



MULTILEVEL CONSTRAINT DECOMPOSITION (MCD)

- dishonest attempt to illustrate MCD as decomposition of ice:



FAS-TYPE COARSE CORRECTIONS

- now add an idea not in Tai (2003)
- suppose:
 - ▶ g^j is solution iterate on current level
 - ▶ down-smoother computes $y^j \in \mathcal{K}^j$
 - ▶ new solution iterate is $\tilde{g}^j = g^j + y^j$
- nonlinear coarse correction needs to restrict residual *and* \tilde{g}^j
 - ▶ **full approximation scheme** (FAS)
 - ▶ in formulas:

$$\begin{aligned}F &= N^j(\tilde{g}^j) - f^j \\g^{j-1} &= R^\bullet \tilde{g}^j \\f^{j-1} &= N^{j-1}(g^{j-1}) - RF\end{aligned}$$

where R^\bullet is state restriction (e.g. injection)

PUT IT TOGETHER: NONLINEAR MCD ALGORITHM

NMCD-VCYCLE(J, w^J, χ^J):

$$g^J = w^J$$

for $j = J$ **downto** $j = 1$

$$\chi^{j-1} = R^\oplus \chi^j$$

$$\phi^j = \chi^j - P\chi^{j-1}$$

$$y^j = 0$$

SMOOTHER^{down}($j, g^j, y^j, N^j, f^j, \phi^j$)

$$F = N^j(g^j + y^j) - f^j$$

$$g^{j-1} = R^\bullet(g^j + y^j)$$

$$f^{j-1} = N^{j-1}(g^{j-1}) - RF$$

$$y^0 = 0$$

SMOOTHER^{coarse}($0, g^0, y^0, N^0, f^0, \chi^0$)

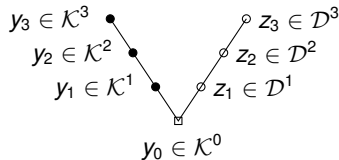
$$z^0 = y^0$$

for $j = 1$ **to** $j = J$

$$z^j = Pz^{j-1} + y^j$$

SMOOTHER^{up}($j, g^j, z^j, N^j, f^j, \chi^j$)

return z^J



up-smoothing corrections
act in larger sets

WHAT IS A GOOD SIA SMOOTHER?

- SIA residual is local
 - ▶ (degenerate) elliptic differential operator
- pointwise smoothers adequate:
 - ▶ projected nonlinear Gauss-Seidel (PNGS)
 - ▶ projected nonlinear Jacobi (PNJacobi; below)

PNJACOBI($j, g^j, y^j, N^j, f^j, b^j, \text{newtonits} = 2, \text{omega} = 0.67$):

for $k = 1, \dots, \text{newtonits}$

$$\rho_p(c) := N^j(g^j + y^j + c\psi_p^j)[\psi_p^j] - f^j[\psi_p^j]$$

$$r_p, \delta_p = \rho_p(0), \rho'_p(0)$$

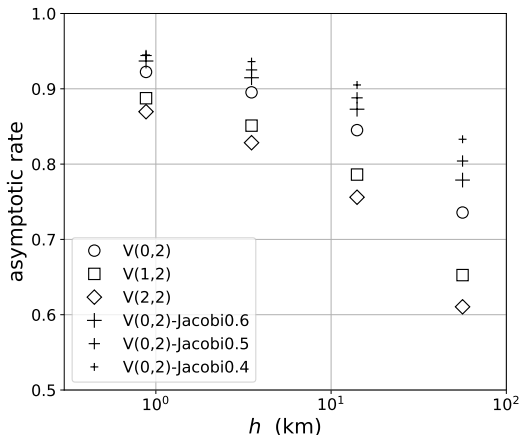
for $p = 1, \dots, m_j$

$$c = \text{POINTUPDATE}(r_p, \delta_p, y_p, b_p, f^j[\psi_p^j])$$

$$y_p \leftarrow y_p + \text{omega } c$$

SIA RESULTS

- compare PNGS and PNJacobi smoothers
- GMG v-cycle factors < 1 for SIA
 - ▶ up-smoothing preferred, thus $V(0,2)$
 - ▶ evidence of mesh independence?



WHAT IS A GOOD STOKES SMOOTHER?

- Stokes residual non-local!
- I have only beginnings of smoother ideas

- goal: GMG for implicit glacier geometry evolution
- much of the tool-chain exists:
 - ▶ backward Euler or other stiff scheme
 - ▶ NCP or VI for free-boundary problem at each time step
 - ▶ nonlinear MCD solution of the VI
 - a form of GMG
- mostly implemented in Firedrake
 - ▶ extruded mesh
 - ▶ mixed-element, GMG solution of Stokes equations
- outlook for entire approach depends on constructing
a performant smoother for Stokes dynamics
- I'm kind of stuck, and seeking help!

extra: REFERENCES

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extra: COMPLETE STRONG FORM

- solve one step of backward Euler for s, \mathbf{u}, p
- system of NCP coupled to Stokes problem:

$$\begin{aligned} s - b &\geq 0 && \text{on } \Omega \\ s - \Delta t \mathbf{u}|_s \cdot \mathbf{n}_s - (s_0 + \Delta t a) &\geq 0 && \text{"} \\ (s - b)(s - \Delta t \mathbf{u}|_s \cdot \mathbf{n}_s - (s_0 + \Delta t a)) &= 0 && \text{"} \\ -\nabla \cdot (2\nu_\epsilon(D\mathbf{u})D\mathbf{u}) + \nabla p - \rho_i \mathbf{g} &= \mathbf{0} && \text{on } \Lambda_s \\ \nabla \cdot \mathbf{u} &= 0 && \text{"} \\ \mathbf{u} &= \mathbf{0} && \text{on } \Gamma_0 \\ (2\nu_\epsilon(D\mathbf{u})D\mathbf{u} - pl) \mathbf{n} &= \mathbf{0} && \text{on } \partial\Lambda_s \setminus \Gamma_0 \end{aligned}$$

- ▶ with regularized Glen-law effective viscosity ($p = \frac{1}{n} + 1$):

$$\nu_\epsilon(D\mathbf{u}) = \frac{\Gamma}{2} (|\mathbf{D}\mathbf{u}|^2 + \epsilon D_0^2)^{(p-2)/2}$$