

# What is ... a *variational inequality*?

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what is ... the source of my title?



W H A T I S . . .

a Gröbner Basis?

*Bernd Sturmfels*



W H A T I S . . .

a Quasi-morphism?

*D. Kotschick*



W H A T I S . . .

a Random Matrix?

*Persi Diaconis*



W H A T I S . . .

a Systole?

*Marcel Berger*

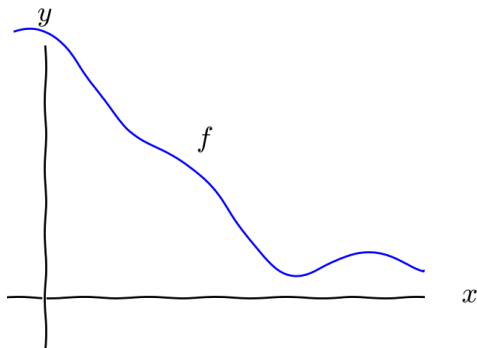
# Outline

problems you can write as variational inequalities

obstacle problem example

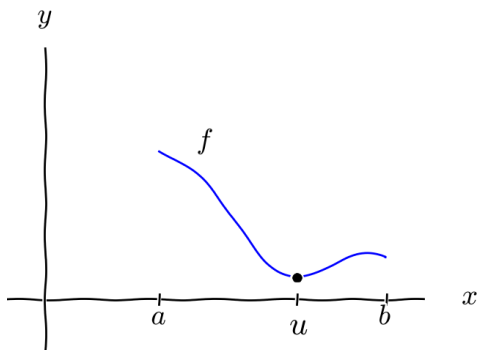
three variational inequalities for glaciers

# calculus I problem



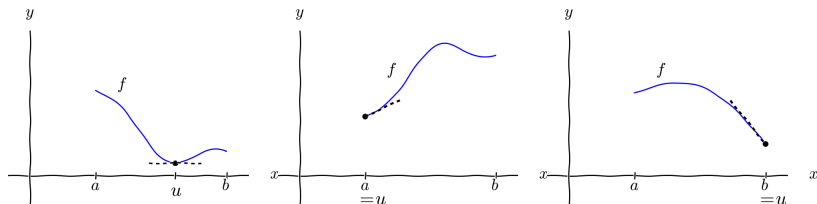
- suppose you have a smooth function
- and you want to minimize it

# calculus I problem



- suppose you have a smooth function *on a closed, bounded interval*
- and you want to minimize it

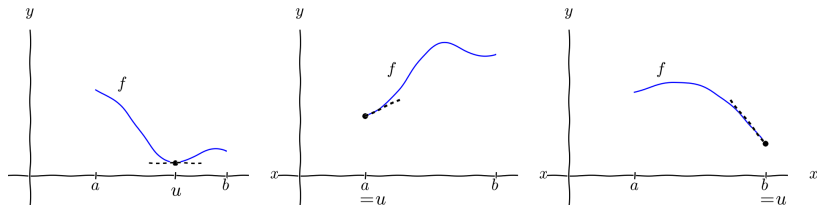
# calculus I problem



because  $f$  is smooth, you can say about the minimizer  $u$  that:

- if  $a < u < b$  then  $f'(u) = 0$  *or*
- if  $u = a$  then  $f'(u) \geq 0$  *or*
- if  $u = b$  then  $f'(u) \leq 0$

# calculus I problem



because  $f$  is smooth, you can say about the minimizer  $u$  that:  
the **variational inequality** applies,

$$f'(u)(v - u) \geq 0 \quad \text{for all } v \in [a, b]$$

## what is a *variational inequality*?

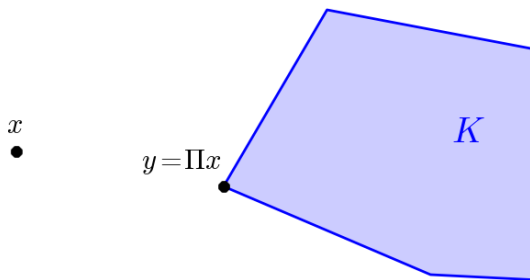
1. cute way to rewrite a calc problem in  $\mathbb{R}^1$ ;  $\min f$  solves:

$$u : \quad f'(u)(v - u) \geq 0 \quad \text{for all } v \in [a, b]$$

the above is a *necessary condition* only



# projection onto a closed, convex set $K \subset \mathcal{H}$



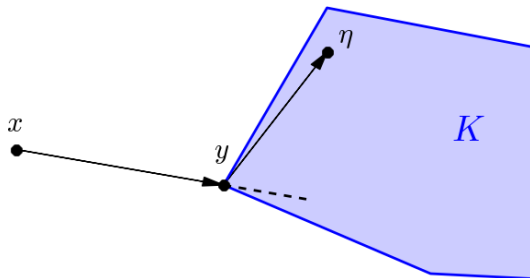
Suppose  $K \subset \mathbb{R}^n$  is closed and convex.  
(Or  $K \subset \mathcal{H}$  closed and convex, where  $\mathcal{H}$  is a Hilbert space.)

Definition. Given  $x \in \mathbb{R}^n$  (or  $x \in \mathcal{H}$ ), the unique minimizer

$$y = \min_{z \in K} \|x - z\|$$

is the *projection* of  $x$  onto  $K$ , written  $y = \Pi x$ .

# projection onto a closed, convex set $K \subset \mathcal{H}$



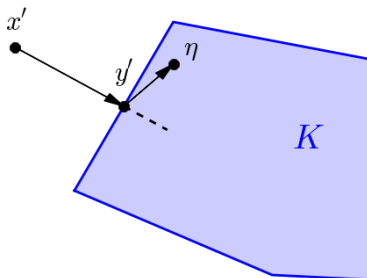
Theorem.

$$y = \Pi x \quad \Longleftrightarrow \quad (y - x) \cdot (\eta - y) \geq 0 \quad \text{for all } \eta \in K$$

This is also a **variational inequality**.

*Idea.* The angle between  $y - x$  and  $\eta - y$  is  $\leq 90^\circ$  for all  $\eta \in K$ .

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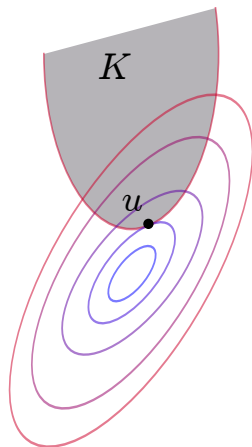
2. dot-product for projection on a closed, convex  $K \subset \mathcal{H}$ :

$$y = \Pi x : \quad (y - x) \cdot (\eta - y) \geq 0 \quad \text{for all } \eta \in K$$

minimize  $f$  on convex  $K \subset \mathbb{R}^n$

Consider a mainstream math problem:

- let  $K \subset \mathbb{R}^n$  be convex
- let  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  be smooth ( $C^1$ )
- find  $u \in K$  so that  $f$  is minimum?



# minimize $f$ on convex $K \subset \mathbb{R}^n$

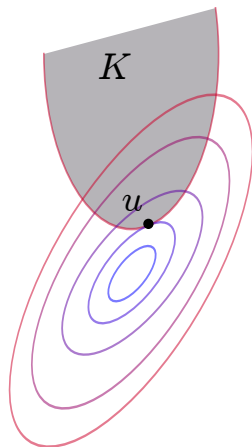
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- find  $u \in K$  so that  $f$  is minimum?

Claim: if  $u \in K$  minimizes  $f$  then  $u$  solves

$$\nabla f(u) \cdot (v - u) \geq 0 \quad \text{for all } v \in K$$

which is a **variational inequality**



minimize  $f$  on convex  $K \subset \mathbb{R}^n$

*Theorem.*  $K \subset \mathbb{R}^n$  convex.  $f : \mathbb{R}^n \rightarrow \mathbb{R}$  smooth.

If  $u \in K$  minimizes  $f$  then

$$\nabla f(u) \cdot (v - u) \geq 0 \quad \text{for all } v \in K.$$

*Proof.* If  $0 \leq t \leq 1$  then

$$(1 - t)u + tv \in K$$

But  $t = 0$  is minimizer of

$$g(t) = f((1 - t)u + tv)$$

so by the “calculus I problem”,  $g'(0) \geq 0$ . By chain rule

$$g'(t) = (\nabla f)((1 - t)u + tv) \cdot (-u + v)$$

so  $g'(0) = (\nabla f)(u) \cdot (v - u) \geq 0$ .



minimize  $f$  on convex  $K \subset \mathbb{R}^n$

How about existence of  $u$ ?

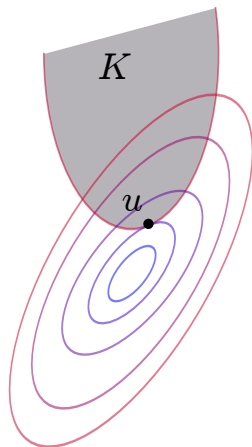
It happens if either:

- $K$  is compact
- $K$  is closed and  $f$  is coercive

How about uniqueness of  $u$ ?

It happens if:

- $f$  is strictly convex





## what is a *variational inequality*?

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3. rewriting of “ $\min f$  on convex  $K \subset \mathbb{R}^n$ ”:

$$u : \quad \nabla f(u) \cdot (v - u) \geq 0 \quad \text{for all } v \in K$$

solve  $F(x) = 0$  on  $\mathbb{R}^n$

Consider another mainstream applied math problem:

- Given continuous function  $F : \mathbb{R}^n \rightarrow \mathbb{R}^n$ , find  $x \in \mathbb{R}^n$  so that

$$F(x) = 0.$$

- That is, solve  $n$  nonlinear equations in  $n$  unknowns.

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$$F(x) = 0.$$

- That is, solve  $n$  nonlinear equations in  $n$  unknowns.
- *No one has anything positive to say about this problem:*
  - no guarantee of existence
  - no guarantee of uniqueness
  - no effective theory of approximation

solve  $F(x) = 0$  on compact, convex  $K \subset \mathbb{R}^n$

But we can change the problem minimally, and have something positive to say:

- Assume  $K \subset \mathbb{R}^n$  is compact and convex.
- Seek  $x \in K$  so that  $F(x) = 0$ .
- Theorem. There is  $x \in K$  so that

$$F(x) \cdot (y - x) \geq 0 \quad \text{for all } y \in K$$

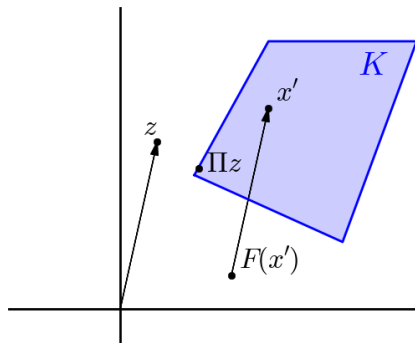
which is a **variational inequality**

solve  $F(x) = 0$  on compact, convex  $K \subset \mathbb{R}^n$

Theorem. There is  $x \in K$  so that

$$F(x) \cdot (y - x) \geq 0 \quad \text{for all } x \in K.$$

Proof.  $x' \mapsto \Pi(x' - F(x'))$ , as map on  $K$ , has a fixed point.



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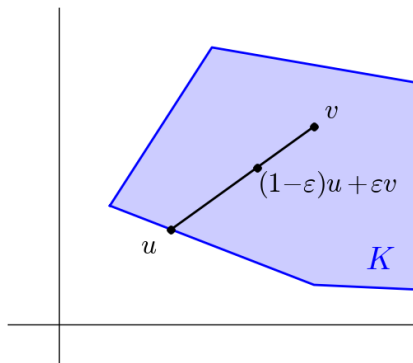
4. gets existence for *any* continuous nonlinear eqns on compact, convex  $K \subset \mathbb{R}^n$ :

$$x : \quad F(x) \cdot (y - x) \geq 0 \quad \text{for all } y \in K$$

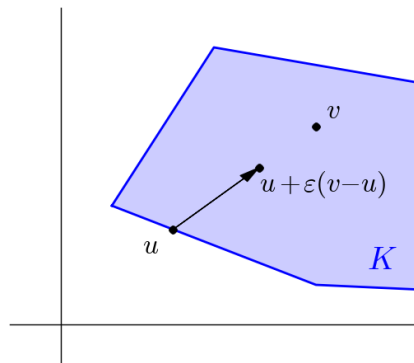
## on convex $K$

if  $K$  is convex and  $u, v \in K$  and  $0 \leq \varepsilon \leq 1$  then

$$(1 - \epsilon)u + \epsilon v = u + \epsilon(v - u) \in K$$



$(1 - \epsilon)u + \epsilon v$  as a linear combination in  $K$



$u + \epsilon(v - u)$  as a vector from  $u$  directed into  $K$

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problems you can write as variational inequalities

obstacle problem example

three variational inequalities for glaciers



## elastic membrane over obstacle

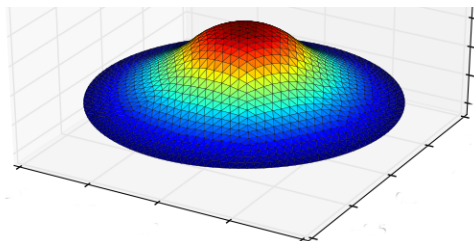
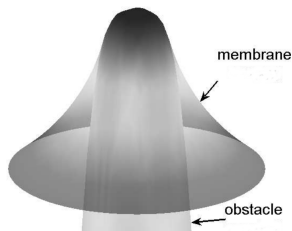
- elastic membrane  $z = u(x, y)$  minimizes energy

$$J[v] = \int_{\Omega} \frac{1}{2} |\nabla v|^2 - f v$$

where  $f$  is upward force on the membrane

- if surface  $v(x, y)$  is above an obstacle  $\psi(x, y)$  then it's in convex set

$$\mathcal{K} = \{v \in H_0^1(\Omega) : v \geq \psi\}$$



# variational inequality for obstacle problem

- if  $u \in \mathcal{K}$  is minimizer and if  $v \in \mathcal{K}$  and if  $0 \leq \epsilon \leq 1$  then

$$\begin{aligned} 0 &\leq J[u + \epsilon(v - u)] - J[u] \\ &= \epsilon \int_{\Omega} \nabla u \cdot \nabla(v - u) - f(v - u) + \epsilon^2 \int_{\Omega} |\nabla(v - u)|^2 \end{aligned}$$

- thus as  $\epsilon \rightarrow 0$ , we know that  $u \in \mathcal{K}$  satisfies

$$\int_{\Omega} \nabla u \cdot \nabla(v - u) - f(v - u) \geq 0 \quad \forall v \in \mathcal{K}$$

which is the variational inequality formulation

- also written:

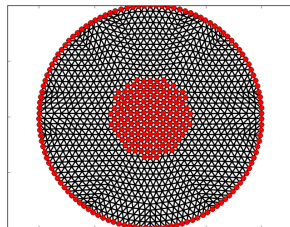
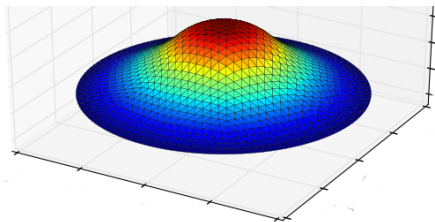
$$\langle \nabla J(u), v - u \rangle \geq 0 \quad \forall v \in \mathcal{K}$$

# PDE for obstacle problem

- where  $u > \psi$ , the variational inequality implies  $-\nabla^2 u = f$ 
  - the standard PDE for an *unobstructed* elastic membrane
  - $-\nabla^2 u = f$  is “Poisson equation”
- an engineer would say

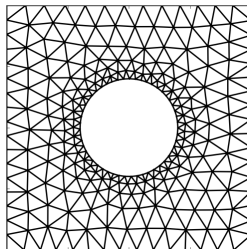
*the membrane  $u(x, y)$  solves  $-\nabla^2 u = f$  except when it is in contact with the obstacle*

- but the set on which the contact happens is *a priori* unknown ...



# finite elements

- the finite element method (FEM) was built on variational *equalities*, i.e. “weak formulations”
- so variational inequalities play well with FEM
- FEM represents (approximates) function spaces on pretty meshes like this:



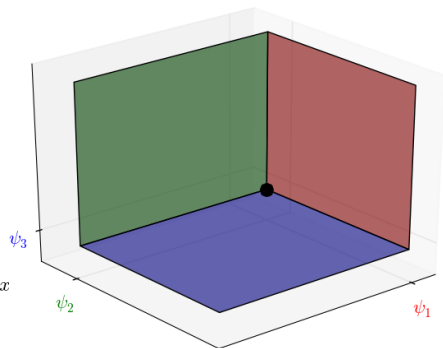
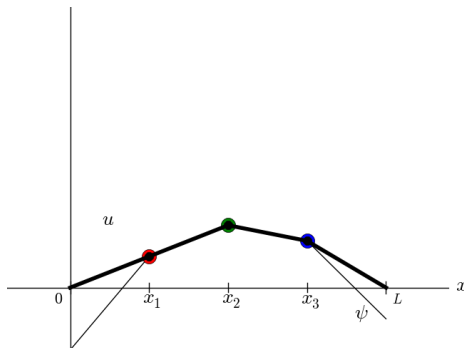
## the 3-point, one-dimensional obstacle problem

- for example, consider a one-dimensional obstacle problem
- with an equally-spaced 3-point mesh
- and a constant force  $f(x) = f_0$
- so  $\Omega = [0, L]$  has mesh points  $\{x_1, x_2, x_3\}$
- the energy is just a quadratic function in  $\mathbb{R}^3$ :

$$\begin{aligned} J[v] &= \int_0^L \frac{1}{2} (v')^2 - f_0 v \\ &\approx \frac{1}{\Delta x} (v_1^2 + v_2^2 + v_3^2 - v_1 v_2 - v_2 v_3) - f_0 \Delta x (v_1 + v_2 + v_3) \end{aligned}$$

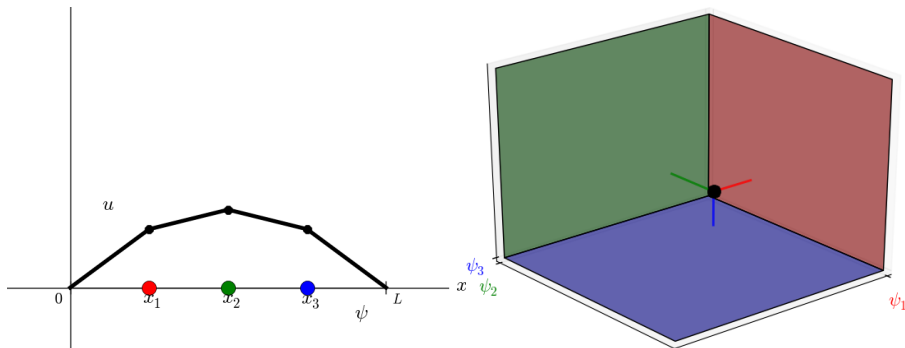
- and  $\psi(x)$  and  $u(x)$  represented by just three values each

# a 3-point case of the obstacle problem



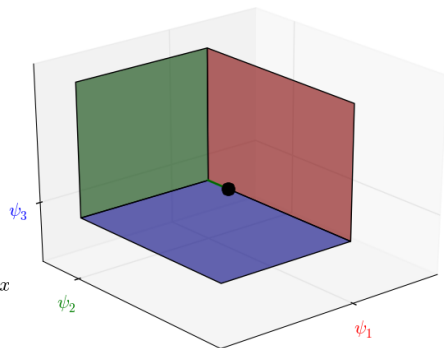
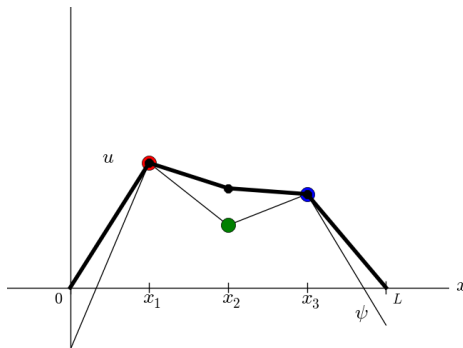
1: zero force  $f_0 = 0$ , one-hump obstacle

# a 3-point case of the obstacle problem



2: upward force  $f_0 > 0$ , flat obstacle

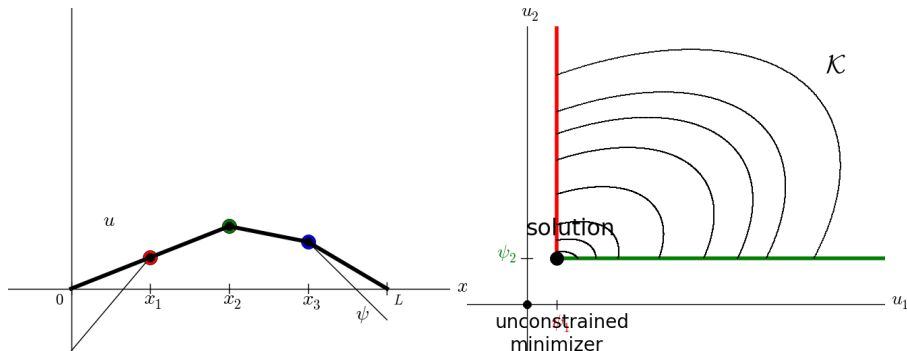
# a 3-point case of the obstacle problem



3: downward force  $f_0 < 0$ , two-peak obstacle

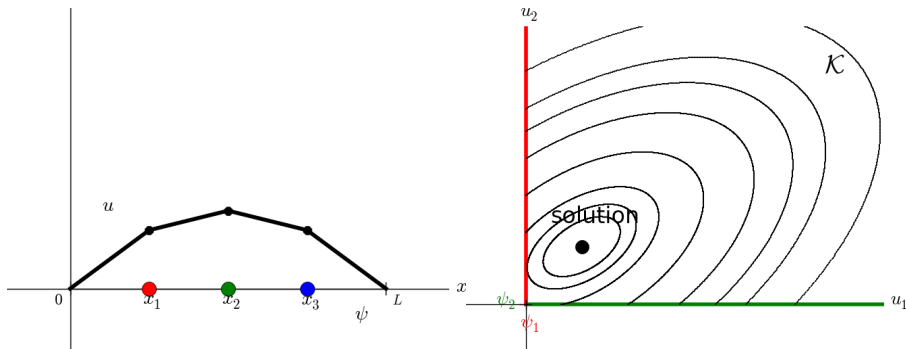


# a 3-point case of obstacle problem



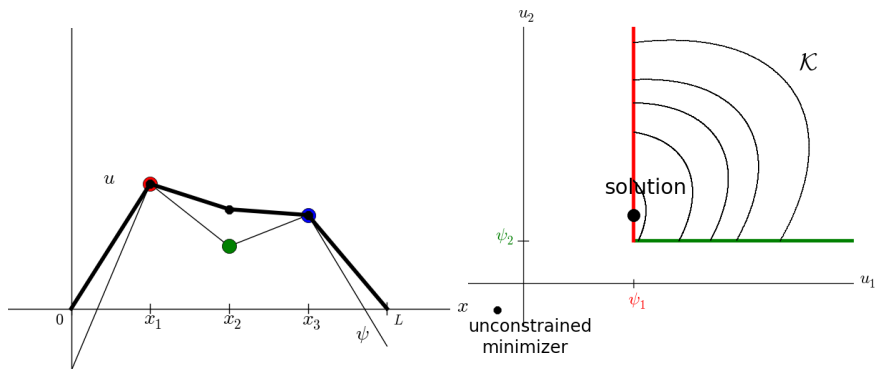
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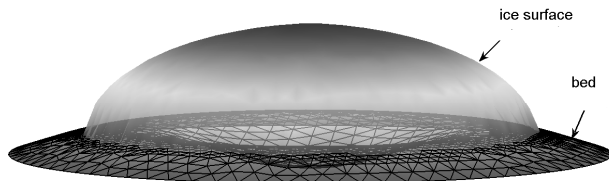
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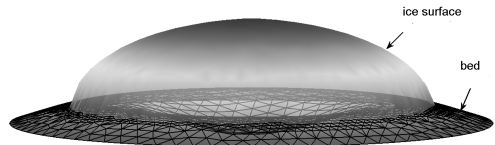
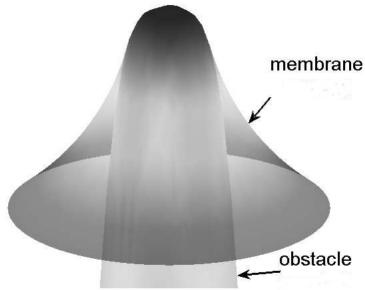
# the steady-climate question for ice sheets

- suppose a steady-state climate
- where it snows some places and melts in others
- the ice flows into areas where there is melting
- questions:
  - what land is covered by ice sheets?
  - how thick are these sheets?



# ice sheet geometry: an obstacle analogy

- ice surface  $s(x, y)$   
 $\sim$  *membrane*
- bedrock  $b(x, y)$   
 $\sim$  *obstacle*



## v.i. 1: steady ice sheet surface “equation”

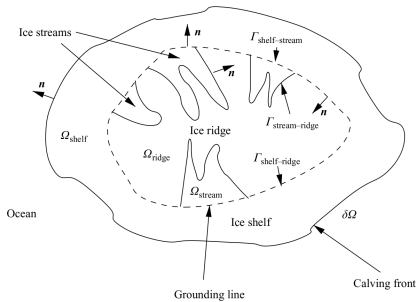
- ice sheet surface equation (so-called “SIA”) applies only on domain where  $s > b$
- let  $h = s - b$ , the ice sheet thickness
- equation applies only where  $s > b \iff h > 0$
- define  $p = n + 1$  where  $n \approx 3$  for shear-thinning ice
- change variables  $h = u^{(p-1)/(2p)}$
- define convex set  $\mathcal{K} = \{v \in W_0^{1,p}(\Omega), v \geq 0\}$
- Theorem (Jouvet-Bueler 2012). There is  $u \in \mathcal{K}$  solving the *steady transformed SIA*,

$$\int_{\Omega} (\mu |\nabla u - \Phi(u)|^{p-2} (\nabla u - \Phi(u))) \cdot \nabla (v - u) \geq \int_{\Omega} \alpha(u) (v - u)$$

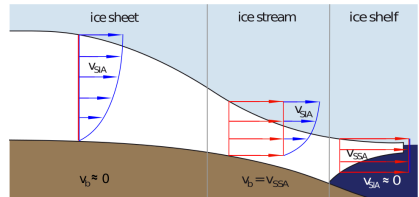
for all  $v \in \mathcal{K}$

# marine ice sheets: overview

- marine ice sheets are full of free boundaries:
  - boundary between floating (“shelf”) and grounded
  - boundary between sliding (“stream”) and not (“sheet”)
  - boundary between wet base and dry base
- the Antarctic ice sheet is a marine ice sheet



cartoon from (Schoof, 2006)

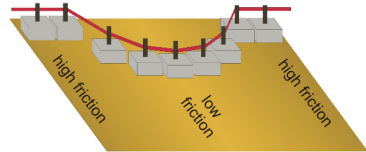
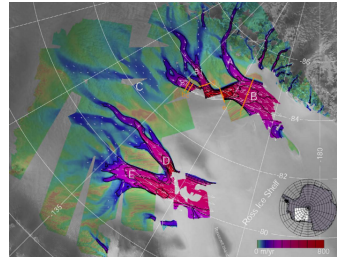


cartoon from (Martin et al., 2011)



# ice stream sliding: an analogy

- ice stream is a viscous membrane with basal stresses
- ice streams emerge where basal resistance is sufficiently low
- a basal resistance model:
  - Coulomb friction, with
  - a yield stress distribution  $\tau_c$
- Schoof's slider analogy



## v.i. 2: ice stream velocity “equation”

- let  $q = 1 + \frac{1}{n}$  where  $n \approx 3$  for shear-thinning ice
- $\mathbf{V}$  is ice stream velocity,  $\mathbf{f} = -\rho g h \nabla s$  is driving stress,  $\mathbf{F}$  is lateral stress along calving front
- Theorem (C. Schoof, 2006). There is unique velocity  $\mathbf{U} = (u, v) \in W^{1,q}(\Omega)$  solving the *coulomb ice stream problem*. It minimizes

$$J[\mathbf{V}] = \int_{\Omega} \frac{2B}{q} h \|\mathbf{V}\|^q + \tau_c |\mathbf{V}| - \mathbf{f} \cdot \mathbf{V} - \int_{\partial\Omega} \mathbf{F} \cdot \mathbf{V}$$

with no constraint

- but  $J[\mathbf{V}]$  is not smooth because of “ $\tau_c |\mathbf{V}|$ ”

## v.i. 2: ice stream velocity “equation”

- Schoof started with a PDE for ice stream velocity (MacAyeal, 1989)
- then derived the variational inequality form:  $\mathbf{U} \in W^{1,q}(\Omega)$  solves

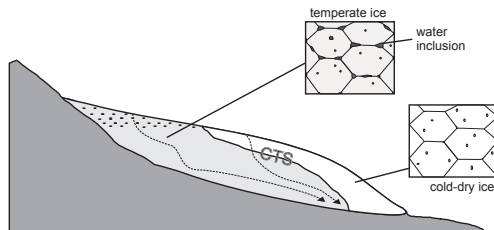
$$\begin{aligned} \int_{\Omega} T_{ij}(\mathbf{U}) D_{ij}(\mathbf{V} - \mathbf{U}) + \tau_c (|\mathbf{V}| - |\mathbf{U}|) - \mathbf{f} \cdot (\mathbf{V} - \mathbf{U}) \\ \geq \int_{\partial\Omega} \mathbf{F} \cdot (\mathbf{V} - \mathbf{U}) \end{aligned}$$

for all  $\mathbf{V} \in W^{1,q}(\Omega)$

- and then got  $J[\mathbf{V}]$  on previous slide

## some glaciers have cold ice

- to a glaciologist, ice is “cold” or “temperate”
  - cold ice has temperature below  $0^{\circ}\text{C}$
  - temperate ice is at  $0^{\circ}\text{C}$ , but with liquid water
- temperature  $u$ , flow velocity  $\mathbf{V}$
- heat flux is  $\mathbf{q} = -k\nabla u + \rho c \mathbf{V}u$ 
  - conductive flux nearly zero in temperate ice ( $\nabla u \approx 0$ )
- viscous dissipation causes heating at rate  $S$



[figures from A. Aschwanden]



# variational inequalities for ice: a summary

- of the three variational inequalities:
  - 1 for the ice sheet surface is *not* a minimization
  - 2 for ice stream sliding is an unconstrained minimization, but of a non-smooth functional
  - 3 for the cold ice is *not* a minimization

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  - 3 for the cold ice is *not* a minimization
- variational inequalities will be used in future glacier and ice sheet problems because of all the free boundaries between different equations
  - when a glaciologist says “this equation describes this . . .” they mean “this equation describes this . . . *wherever the equation can be applied and I can’t generally tell you where that is*”
  - this makes the job of building an ice sheet model harder

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- variational inequalities will be used in future glacier and ice sheet problems because of all the free boundaries between different equations
  - when a glaciologist says “this equation describes this . . .” they mean “this equation describes this . . . *wherever the equation can be applied and I can’t generally tell you where that is*”
  - this makes the job of building an ice sheet model harder
- **and that’s it on variational inequalities for today**