Linking Scales in the Sea Ice System

Kenneth M. Golden Department of Mathematics University of Utah



Institute for Marine and Antarctic Studies, University of Tasmania 13 December 2017

SEA ICE covers ~15% of Earth's ocean surface

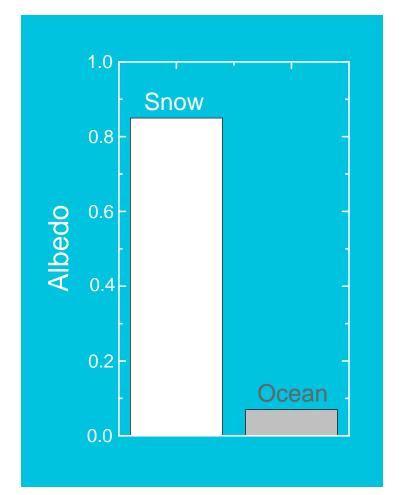
- boundary between ocean and atmosphere
- mediates exchange of heat, gases, momentum
- global ocean circulation
- indicator and agent of climate change

polar ice caps critical to global climate in reflecting incoming solar radiation

white snow and ice reflect



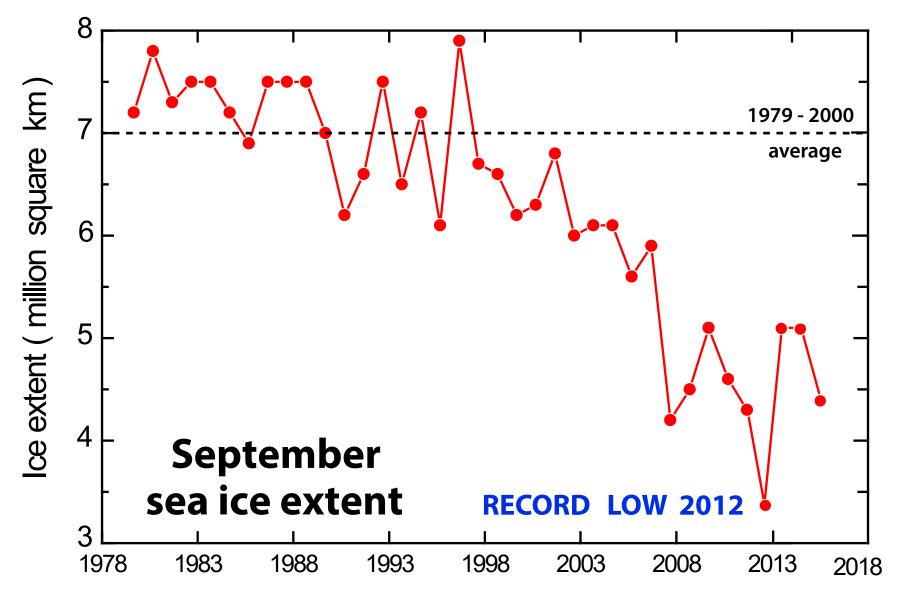




dark water and land absorb

albedo
$$\alpha = \frac{\text{reflected sunlight}}{\text{incident sunlight}}$$

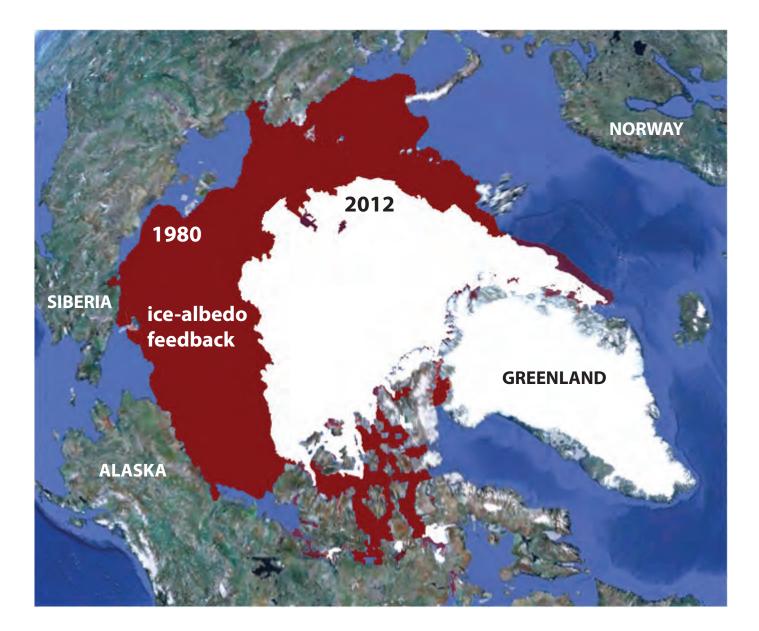
the summer Arctic sea ice pack is melting



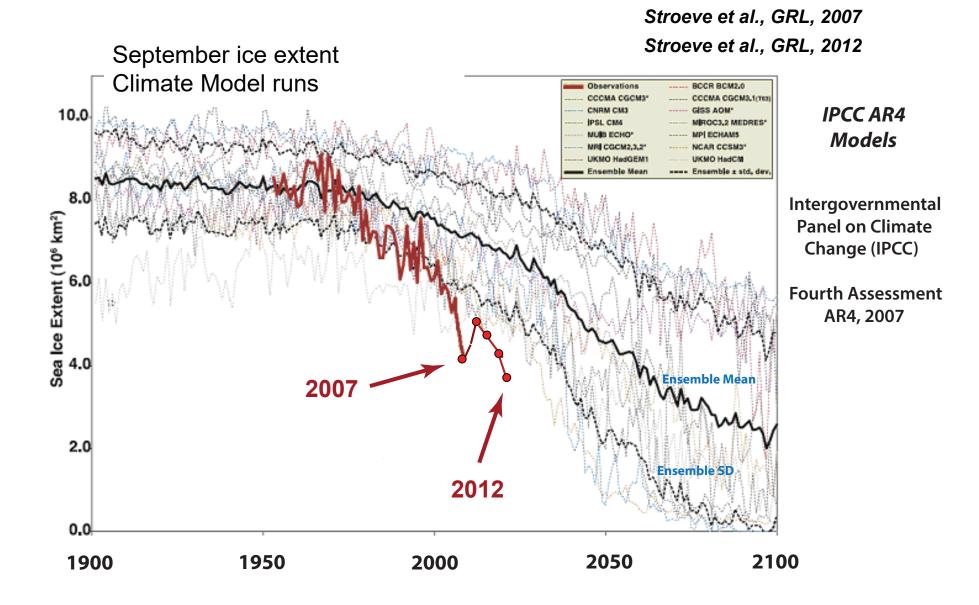
National Snow and Ice Data Center

Change in Arctic Sea Ice Extent

September 1980 -- 7.8 million square kilometers September 2012 -- 3.4 million square kilometers



Arctic sea ice decline: faster than predicted by climate models



challenge

represent sea ice more rigorously in climate models

account for key processes such as melt pond evolution



Impact of melt ponds on Arctic sea ice simulations from 1990 to 2007

Flocco, Schroeder, Feltham, Hunke, JGR Oceans 2012

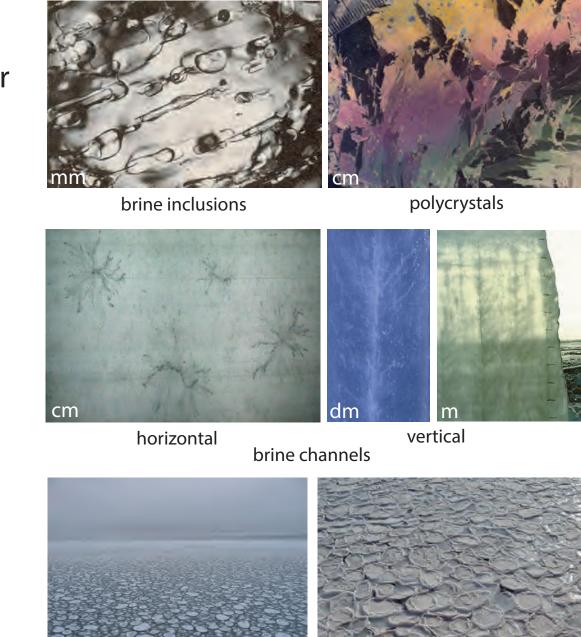
For simulations with ponds September ice volume is nearly 40% lower.

... and other sub-grid scale structures and processes *linkage of scales*

sea ice is a multiscale composite displaying structure over 10 orders of magnitude

0.1 millimeter

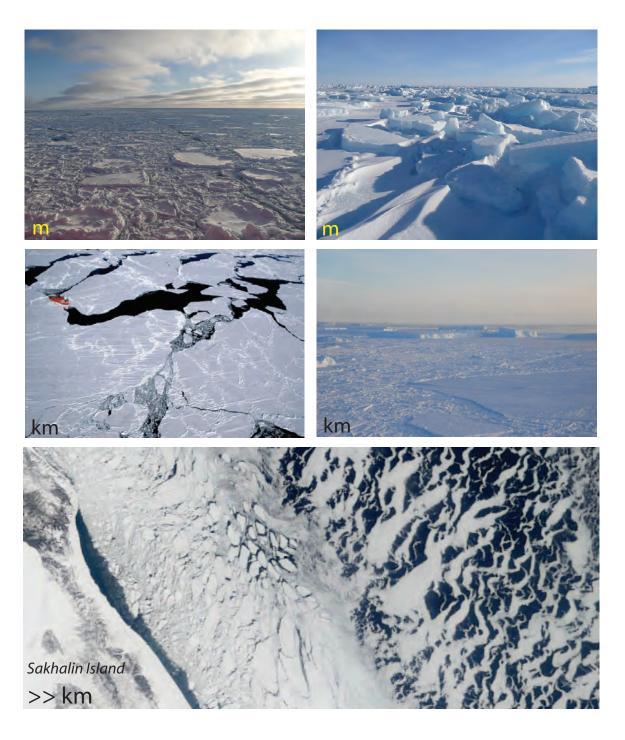
1 meter



pancake ice

1 meter

100 kilometers

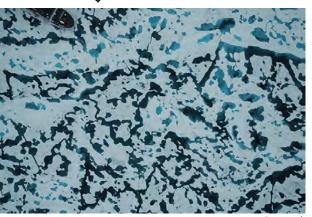




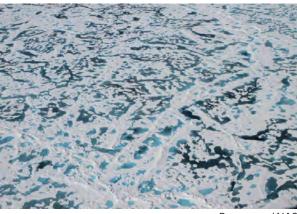
basin scale grid scale albedo

Linking Scales

km scale melt ponds



Perovich



km scale melt ponds

Ramsayer / NASA

Linking

mm scale brine inclusions



Weeks & Assur





meter scale snow topography

Colon

What is this talk about?

Using the mathematics of composite materials and statistical physics to LINK SCALES in the sea ice system ... rigorously compute effective behavior ... to improve climate projections.

HOMOGENIZATION

1. Sea ice microphysics and porous media

fluid flow, diffusion processes, percolation theory

2. EM monitoring of sea ice

Stieltjes integrals, spectral measures, random matrix theory

3. Advection diffusion; polycrystals; waves in the MIZ

integral representations, bounds

- 4. Low order predictors and basin scale homogenization
- 5. Evolution of Arctic melt ponds, fractal geometry

continuum percolation, network and Ising models

critical behavior cross - pollination

sea ice microphysics

fluid transport

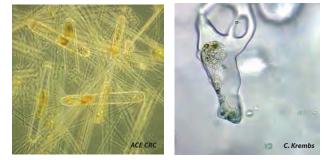
fluid flow through the porous microstructure of sea ice governs key processes in polar climate and ecosystems

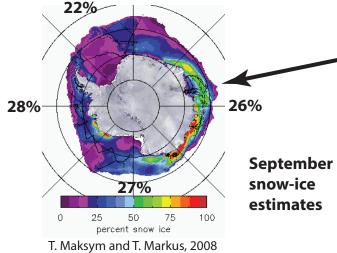
evolution of Arctic melt ponds and sea ice albedo



nutrient flux for algal communities







Antarctic surface flooding and snow-ice formation

evolution of salinity profiles
ocean-ice-air exchanges of heat, CO₂

sea ice ecosystem



sea ice algae support life in the polar oceans

fluid permeability k of a porous medium



porous

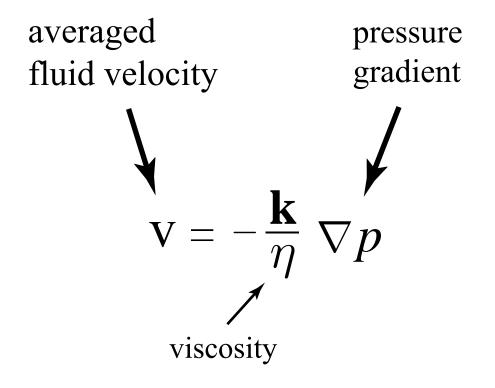
concrete

how much water gets through the sample per unit time?

HOMOGENIZATION

mathematics for analyzing effective behavior of heterogeneous systems

Darcy's Law for slow viscous flow in a porous medium



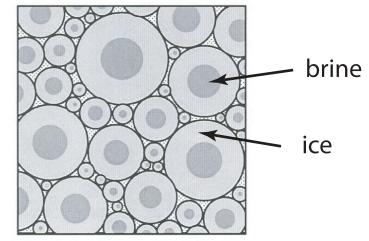
 $\mathbf{k} =$ fluid permeability tensor

PIPE BOUNDS on vertical fluid permeability k

Golden, Heaton, Eicken, Lytle, Mech. Materials 2006 Golden, Eicken, Heaton, Miner, Pringle, Zhu, Geophys. Res. Lett. 2007

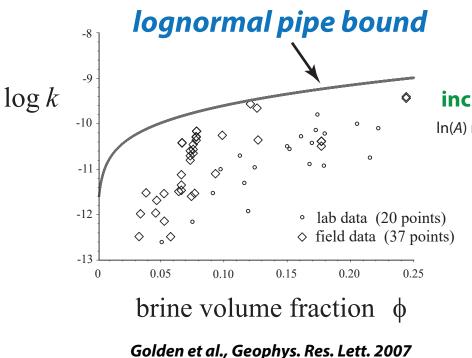
> vertical pipes with appropriate radii maximize k





fluid analog of arithmetic mean upper bound for effective conductivity of composites (Wiener 1912)

optimal coated cylinder geometry



$$k \leq \frac{\phi \langle R^4 \rangle}{8 \langle R^2 \rangle} = \frac{\phi}{8} \langle R^2 \rangle e^{\sigma^2}$$

inclusion cross sectional areas A lognormally distributed

In(A) normally distributed, mean μ (increases with T) variance $\sigma^{_2}(\mbox{Gow and Perovich 96})$

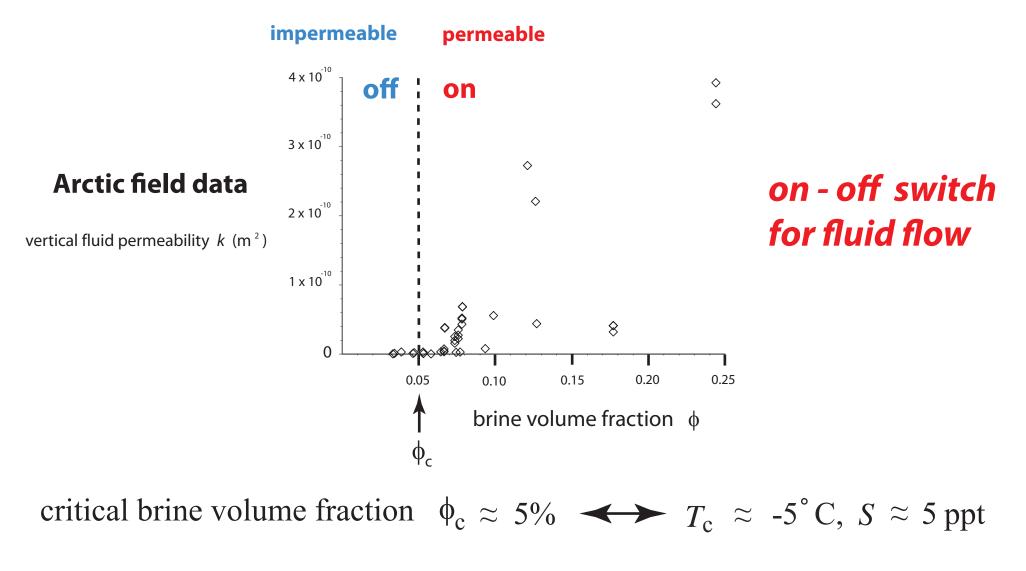
get bounds through variational analyis of **trapping constant** γ for diffusion process in pore space with absorbing BC

Torquato and Pham, PRL 2004

 $\mathbf{k} \leq \gamma^{-1} \mathbf{I}$

for any ergodic porous medium (Torquato 2002, 2004)

Critical behavior of fluid transport in sea ice



RULE OF FIVES

Golden, Ackley, Lytle Science 1998Golden, Eicken, Heaton, Miner, Pringle, Zhu, Geophys. Res. Lett. 2007Pringle, Miner, Eicken, Golden J. Geophys. Res. 2009



sea ice algal communities

D. Thomas 2004

nutrient replenishment controlled by ice permeability

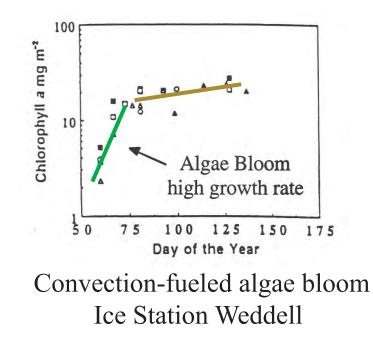
biological activity turns on or off according to *rule of fives*

Golden, Ackley, Lytle

Science 1998

Fritsen, Lytle, Ackley, Sullivan Science 1994

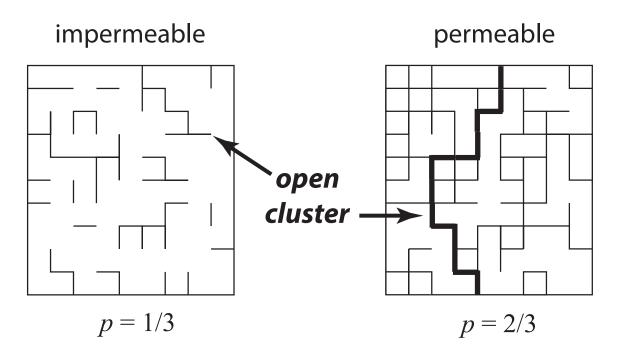
critical behavior of microbial activity



Why is the rule of fives true?

percolation theory

probabilistic theory of connectedness



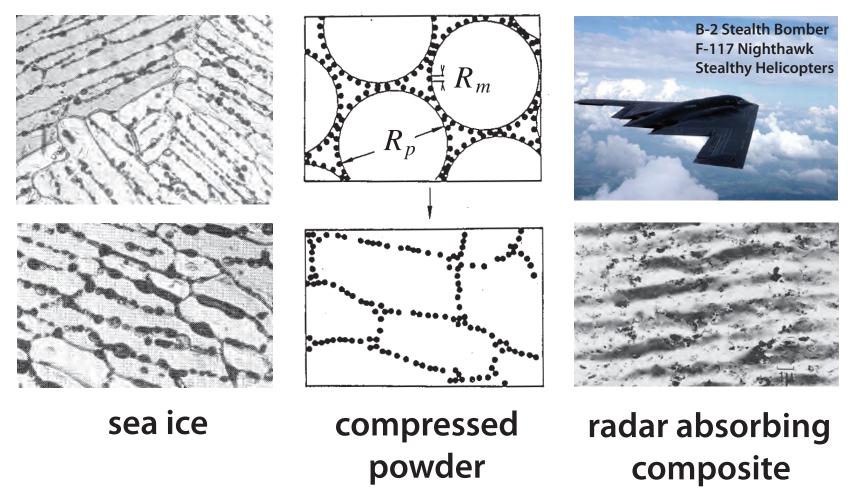
bond \longrightarrow *open with probability p closed with probability 1-p*

percolation threshold $p_c = 1/2$ for d = 2

smallest *p* for which there is an infinite open cluster

Continuum percolation model for *stealthy* materials applied to sea ice microstructure explains **Rule of Fives** and Antarctic data on ice production and algal growth

 $\phi_c \approx 5\%$ Golden, Ackley, Lytle, *Science*, 1998



sea ice is radar absorbing

Thermal evolution of permeability and microstructure in sea ice Golden, Eicken, Heaton, Miner, Pringle, Zhu



rigorous bounds percolation theory hierarchical model network model

field data

X-ray tomography for brine inclusions

unprecedented look at thermal evolution of brine phase and its connectivity

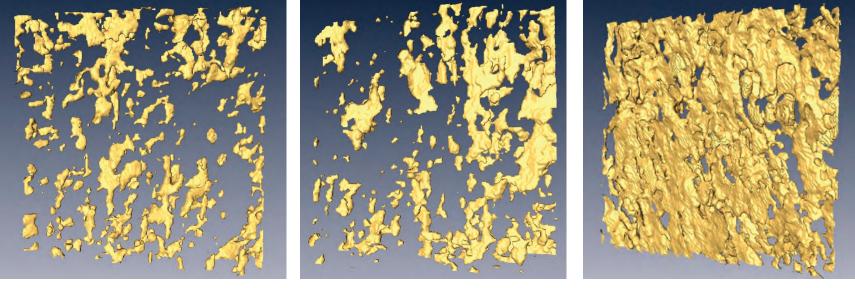
controls

micro-scale

macro-scale processes

brine connectivity (over cm scale)

8 x 8 x 2 mm



-15 °C, $\phi = 0.033$ -6 °C, $\phi = 0.075$ -3 °C, $\phi = 0.143$

X-ray tomography confirms percolation threshold

3-D images 3-D graph ores and throats nodes and edges

analyze graph connectivity as function of temperature and sample size

- use finite size scaling techniques to confirm rule of fives
- order parameter data from a natural material

Pringle, Miner, Eicken, Golden, J. Geophys. Res. 2009

lattice and continuum percolation theories yield:

$$k(\phi) = k_0 (\phi - 0.05)^2 \qquad \text{critical} \\ \text{exponent} \\ k_0 = 3 \times 10^{-8} \text{ m}^2 \qquad t$$

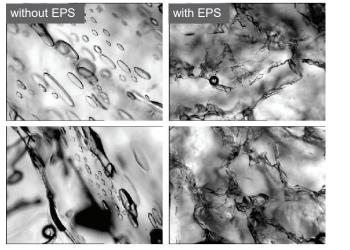
- exponent is UNIVERSAL lattice value $t \approx 2.0$
- sedimentary rocks like sandstones also exhibit universality
- critical path analysis -- developed for electronic hopping conduction -- yields scaling factor k_0

$$y = \log k \xrightarrow{-7}_{-8}_{-10}$$

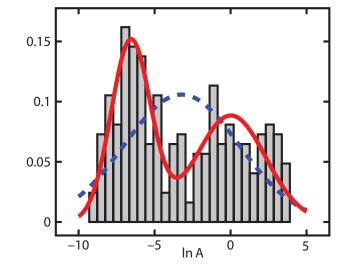
theory: $y = 2 \times -7.5$
 $y = \log k \xrightarrow{-9}_{-10}_{-11}_{-12}_{-13}_{-14}_{-12}_{-13}_{-14}_{-15}_{-2.2 -2}_{-2.2 -2}_{-1.8 -1.6 -1.4 -1.2 -1}_{-1.6 -1.4 -1.2 -1}_{x = \log(\phi - 0.05)}$

Sea ice algae secrete extracellular polymeric substances (EPS) affecting evolution of brine microstructure.

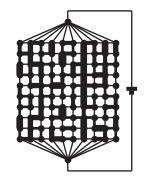
How does EPS affect fluid transport?



Krembs, Eicken, Deming, PNAS 2011



RANDOM PIPE MODEL



 $R_{i-1,j}^{h} \underbrace{\begin{array}{c} R_{i,j}^{v} \\ (i,j) \\ R_{i,j-1}^{h} \end{array}}_{R_{i,j-1}^{v}} R_{i,j}^{h}$

Zhu, Jabini, Golden, Eicken, Morris *Ann. Glac*. 2006

- **Bimodal** lognormal distribution for brine inclusions
- Develop random pipe network model with bimodal distribution;
 Use numerical methods that can handle larger variances in sizes.
- Results predict observed drop in fluid permeability k.
- Rigorous bound on *k* for bimodal distribution of pore sizes

Steffen, Epshteyn, Zhu, Bowler, Deming, Golden Multiscale Modeling and Simulation, in press.

How does the biology affect the physics?

Remote sensing of sea ice



sea ice thickness ice concentration

INVERSE PROBLEM

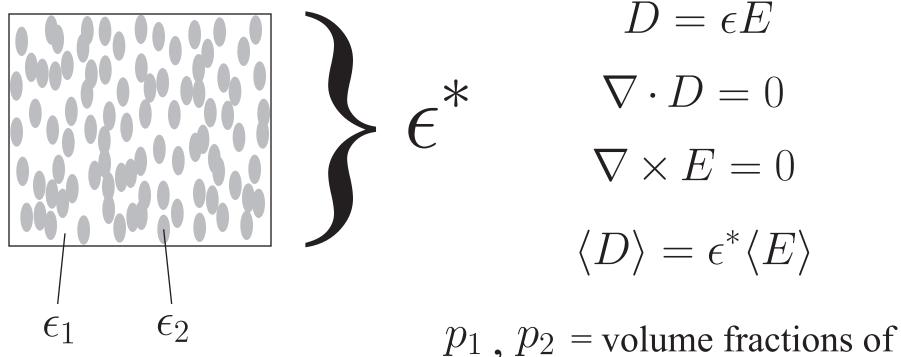
Recover sea ice properties from electromagnetic (EM) data

8*

effective complex permittivity (dielectric constant, conductivity)



brine volume fraction brine inclusion connectivity Effective complex permittivity of a two phase composite in the quasistatic (long wavelength) limit



the components

 $\epsilon^* = \epsilon^* \left(\frac{\epsilon_1}{\epsilon_2} , \text{ composite geometry} \right)$

Theory of Effective Electromagnetic Behavior of Composites analytic continuation method

Forward Homogenization Bergman (1978), Milton (1979), Golden and Papanicolaou (1983) *Theory of Composites*, Milton (2002)

> **composite geometry** (spectral measure μ)



integral representations, rigorous bounds, approximations, etc.

$$F(s) = 1 - \frac{\epsilon^*}{\epsilon_2} = \int_0^1 \frac{d\mu(z)}{s-z} \qquad s = \frac{1}{1 - \epsilon_1/\epsilon_2} \qquad \xrightarrow{\circ} \qquad$$

Inverse Homogenization Cherkaev and Golden (1998), Day and Thorpe (1999), Cherkaev (2001) McPhedran, McKenzie, Milton (1982), *Theory of Composites*, Milton (2002)



recover brine volume fraction, connectivity, etc.

Stieltjes integral representation separates geometry from parameters

$$F(s) = 1 - \frac{\epsilon^*}{\epsilon_2} = \int_0^1 \frac{d\mu(z)}{s-z}$$

spectral measure of self adjoint operator Γ χ
 mass = p₁
 higher moments depend on *n*-point correlations

$$\Gamma = \nabla (-\Delta)^{-1} \nabla \cdot$$

 $\chi = {\rm characteristic \, function} \\ {\rm of \, the \, brine \, phase}$

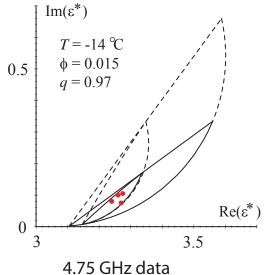
$$E = (s + \Gamma \chi)^{-1} e_k$$

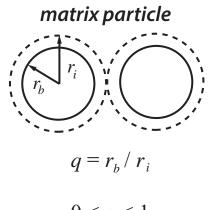
$\Gamma \chi$: microscale \rightarrow macroscale $\Gamma \chi$ *links scales*

forward and inverse bounds on the complex permittivity of sea ice







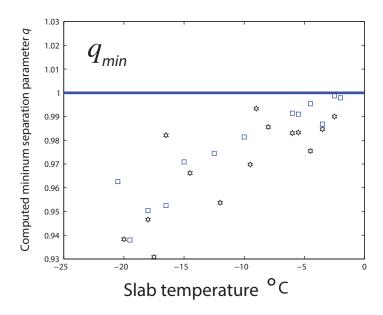


0 < q < 1

Golden 1995, 1997 Bruno 1991

inverse bounds and recovery of brine porosity

Gully, Backstrom, Eicken, Golden *Physica B, 2007*



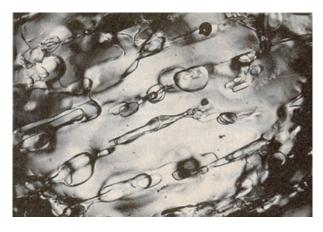
inversion for brine inclusion separations in sea ice from measurements of effective complex permittivity ϵ^*

rigorous inverse bound on spectral gap

construct algebraic curves which bound admissible region in (p,q)-space

Orum, Cherkaev, Golden *Proc. Roy. Soc. A, 2012*

SEA ICE

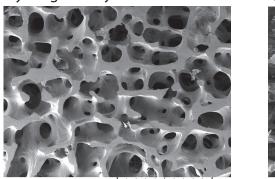


young healthy trabecular bone



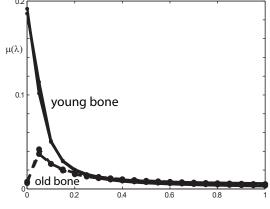
spectral characterization of porous microstructures in human bone

reconstruct spectral measures from complex permittivity data



m





use regularized inversion scheme

apply spectral measure analysis of brine connectivity and spectral inversion to electromagnetic monitoring of osteoporosis

Golden, Murphy, Cherkaev, J. Biomechanics 2011

the math doesn't care if it's sea ice or bone!

direct calculation of spectral measure

- 1. Discretization of composite microstructure gives lattice of 1's and 0's (random resistor network).
- 2. The fundamental operator $\chi\Gamma\chi$ becomes a random matrix depending only on the composite geometry.
- 3. Compute the eigenvalues λ_i and eigenvectors of $\chi \Gamma \chi$ with inner product weights α_i

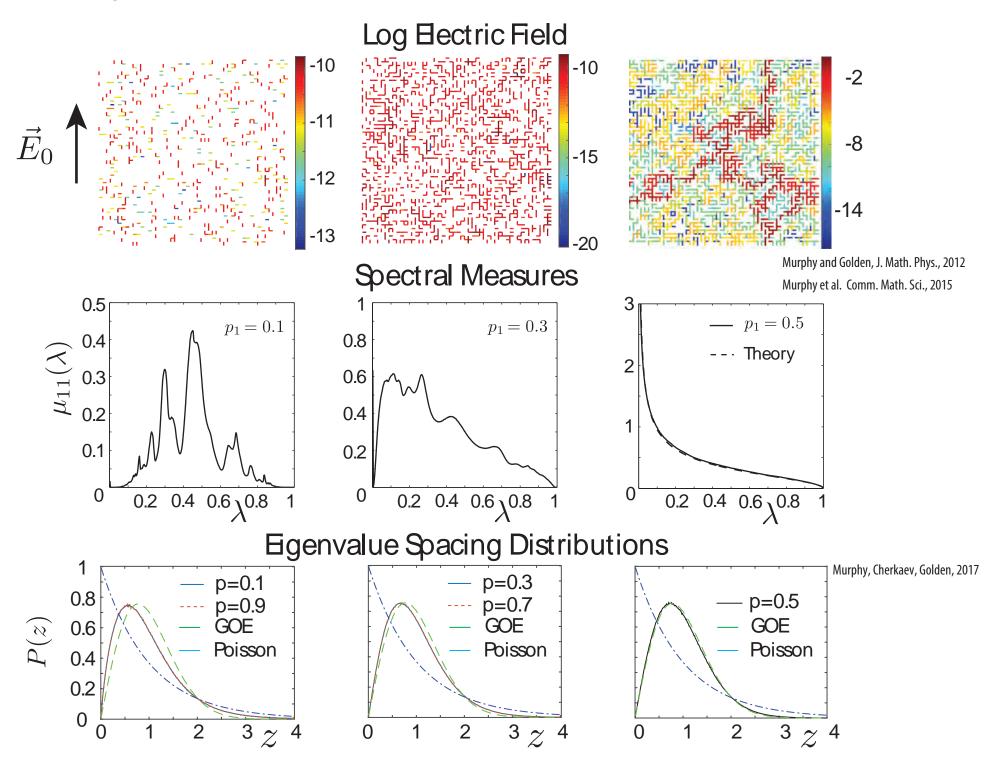
$$\mu(\lambda) = \sum_{i} \alpha_{i} \delta(\lambda - \lambda_{i})$$

Dirac point measure (Dirac delta)

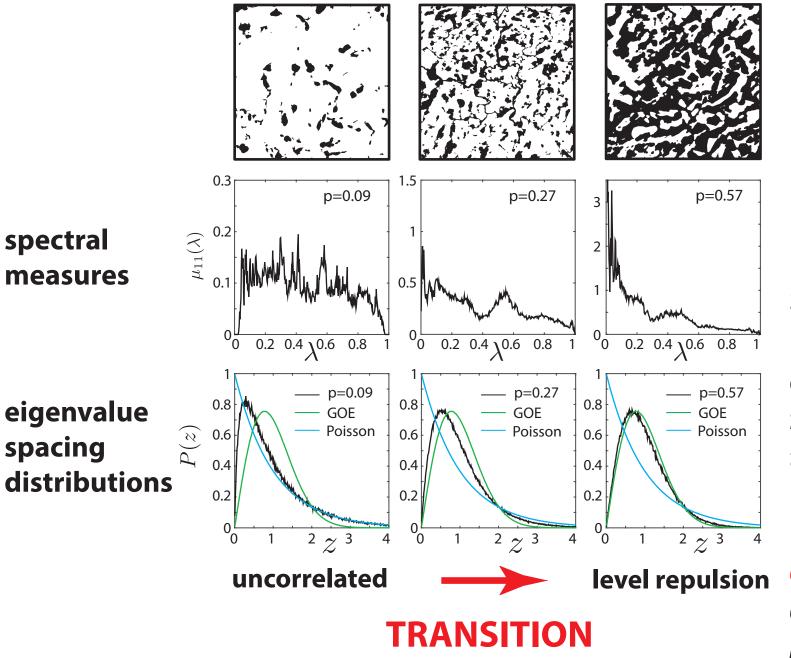
earlier studies of spectral measures

Day and Thorpe 1996 Helsing, McPhedran, Milton 2011

Spectral statistics for 2D random resistor network

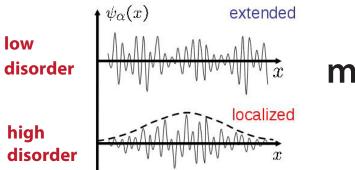


Spectral computations for Arctic melt ponds



Ben Murphy Elena Cherkaev Ken Golden 2017

eigenvalue statistics for transport tend toward the UNIVERSAL Wigner-Dyson distribution as the "conducting" phase percolates



metal / insulator transition localization

Anderson 1958 Mott 1949 Shklovshii et al 1993 Evangelou 1992

Anderson transition in wave physics: quantum, optics, acoustics, water waves, ...

we find a surprising analog

Anderson transition for classical transport in composites

Murphy, Cherkaev, Golden Phys. Rev. Lett. 2017

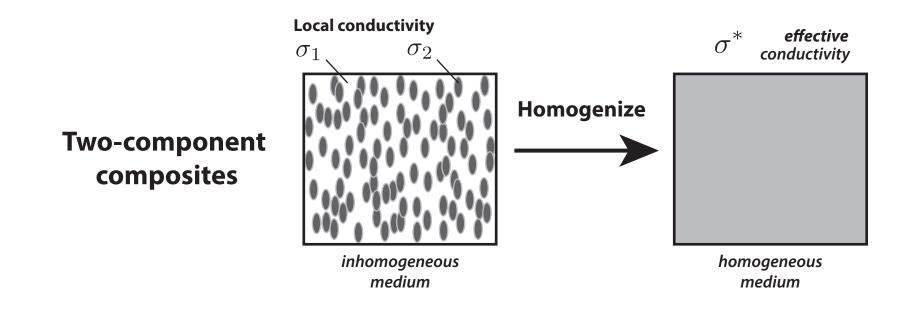




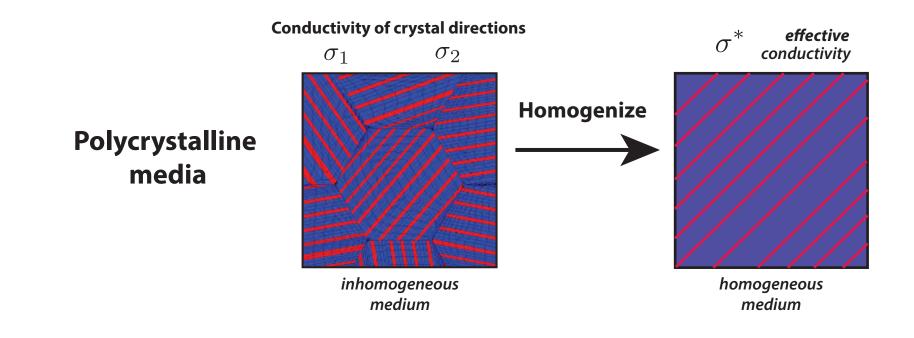
transition to universal eigenvalue statistics (GOE) extended states, mobility edges

-- but without wave interference or scattering effects ! --

Homogenization for composite materials



Find the homogeneous medium which behaves macroscopically the same as the inhomogeneous medium



Bounds on the complex permittivity of polycrystalline materials by analytic continuation

> Adam Gully, Joyce Lin, Elena Cherkaev, Ken Golden

 Stieltjes integral representation for effective complex permittivity

Milton (1981, 2002), Barabash and Stroud (1999), ...

- Forward and inverse bounds
- Applied to sea ice using two-scale homogenization
- Inverse bounds give method for distinguishing ice types using remote sensing techniques





Proc. Roy. Soc. A 8 Feb 2015

ISSN 1364-5021 | Volume 471 | Issue 2174 | 8 February 2015

PROCEEDINGS A

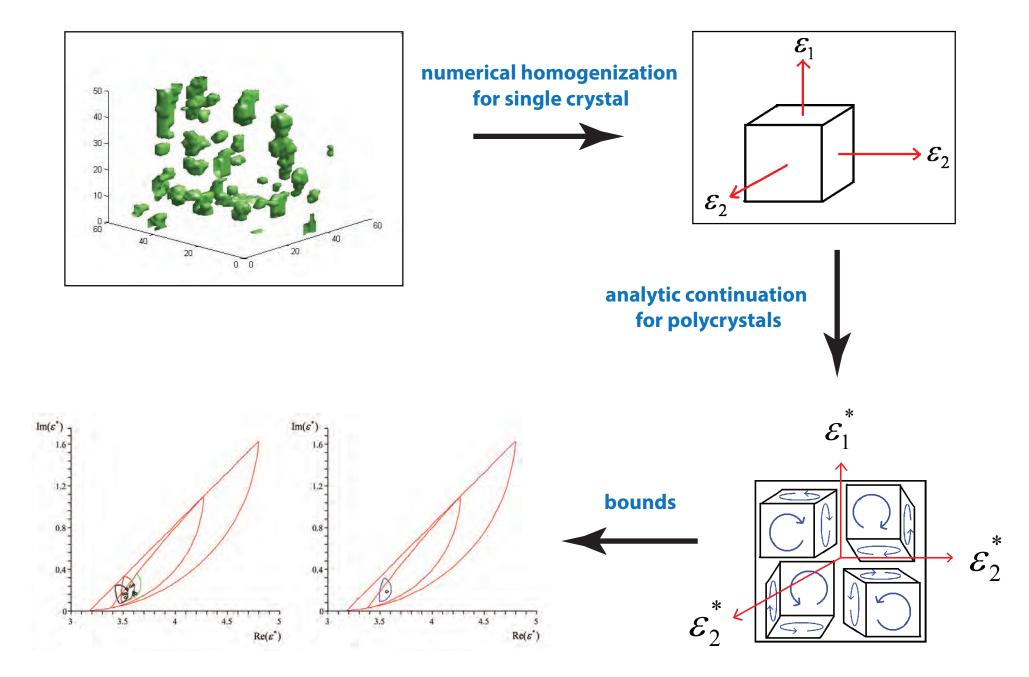


An invited review commemorating 350 years of scientific publishing at the Royal Society

A method to distinguish between different types of sea ice using remote sensing techniques A computer model to determine how a human should walk so as to expend the least energy



two scale homogenization for polycrystalline sea ice



Gully, Lin, Cherkaev, Golden, Proc. Roy. Soc. A (and cover) 2015

advection enhanced diffusion

effective diffusivity

tracers, buoys diffusing in ocean eddies diffusion of pollutants in atmosphere salt and heat transport in ocean heat transport in sea ice with convection

advection diffusion equation with a velocity field $\,ec u\,$

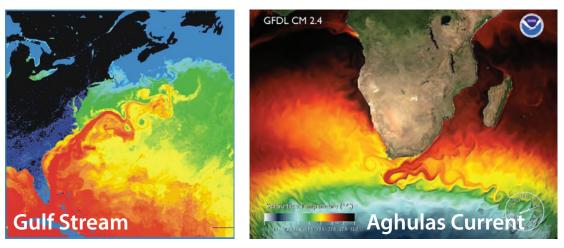
$$\frac{\partial T}{\partial t} + \vec{u} \cdot \vec{\nabla}T = \kappa_0 \Delta T$$
$$\vec{\nabla} \cdot \vec{u} = 0$$
$$homogenize$$
$$\frac{\partial \overline{T}}{\partial t} = \kappa^* \Delta \overline{T}$$

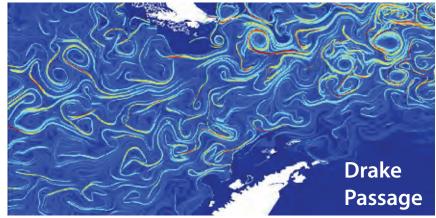
κ^{*} effective diffusivity

Stieltjes integral for κ^* with spectral measure

Avellaneda and Majda, PRL 89, CMP 91

Murphy, Cherkaev, Xin, Zhu, Golden, *Ann. Math. Sci. Appl.* 2017 Murphy, Cherkaev, Zhu, Xin, Golden, 2017

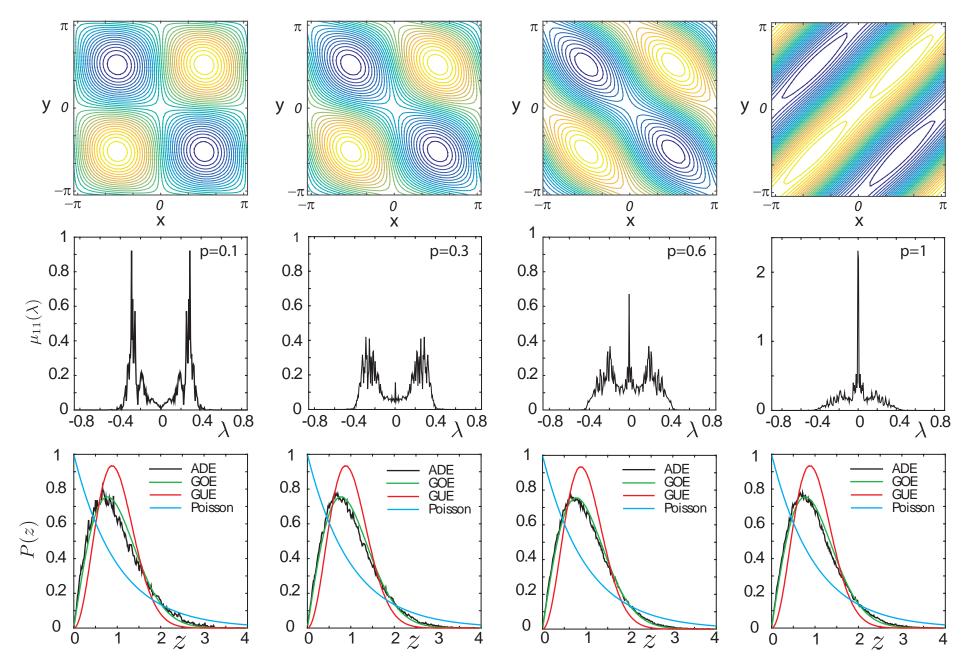






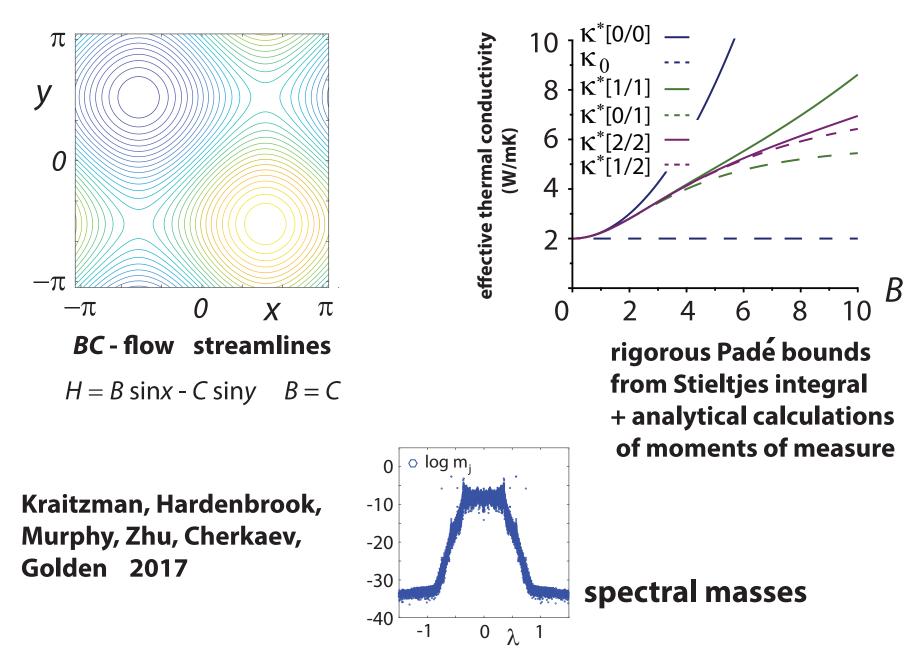
Spectral measures and eigenvalue spacings for cat's eye flow

 $H(x,y) = sin(x) sin(y) + A cos(x) cos(y), \quad A \sim U(-p,p)$



Murphy, Cherkaev, Xin, Golden, 2017

Convection - enhanced thermal conductivity of sea ice w/ BC - flow



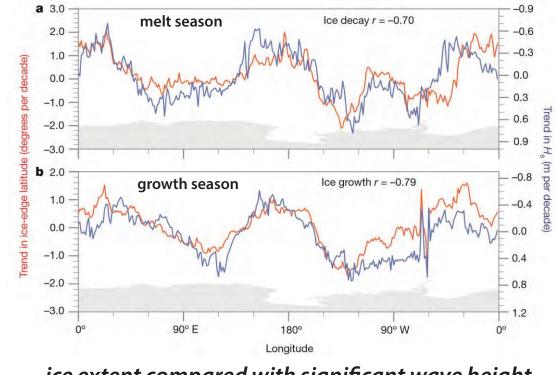
Murphy, Cherkaev, Zhu, Xin, Golden 2017

Storm-induced sea-ice breakup and the implications for ice extent

Kohout et al., Nature 2014

- during three large-wave events, significant wave heights did not decay exponentially, enabling large waves to persist deep into the pack ice.
- Iarge waves break sea ice much farther from the ice edge than would be predicted by the commonly assumed exponential decay

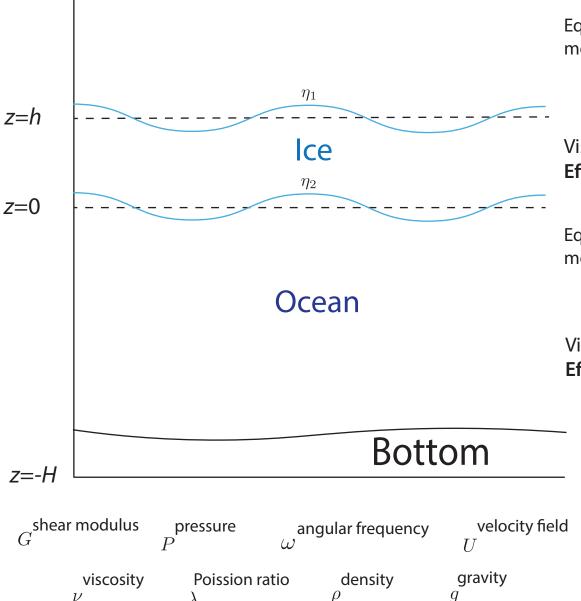




ice extent compared with significant wave height

Waves have strong influence on both the floe size distribution and ice extent.

Two Layer Models and Effective Parameters



 ν

Viscous fluid layer (Keller 1998) Effective Viscosity ν

Equations of $\frac{\partial U}{\partial t} = -\frac{1}{\rho}\nabla P + \nu\nabla^2 U + g$

Viscoelastic fluid layer (Wang-Shen 2010) Effective Complex Viscosity $\nu_e = \nu + iG/\rho\omega$

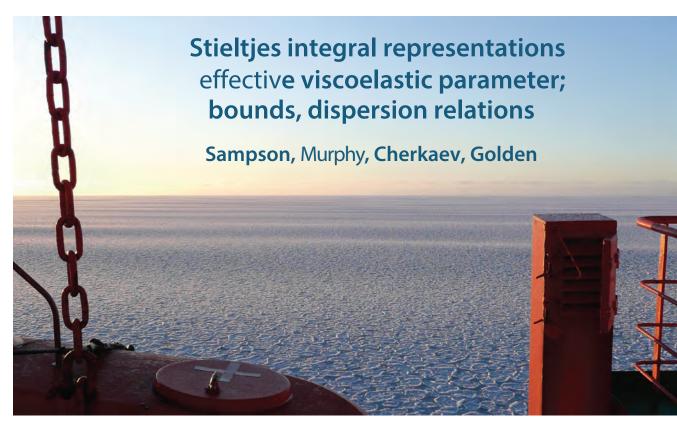
Equations of $\frac{\partial U}{\partial t} = -\frac{1}{\rho}\nabla P + \nu_e \nabla^2 U + g$

Viscoelastic thin beam (Mosig et al. 2015) Effective Complex Shear Modulus $G_v = G - i\omega\rho\nu$

Stieltjes integral representation for effective complex viscoelastic parameter; bounds

Sampson, Murphy, Cherkaev, Golden 2017

wave propagation in the marginal ice zone





Stieltjes Integral Representation for Complex Viscoelasticity

$$\nabla \cdot \sigma = 0 \qquad \sigma_{ij} = C_{ijkl}\epsilon_{kl} \qquad \langle \sigma_{ij} \rangle = C_{ijkl}^* \langle \epsilon_{kl} \rangle \qquad \text{Strain Field}$$

$$\text{local} \qquad C_{ijkl} = (\nu_1 \chi + (1 - \chi)\nu_2)\lambda_s \qquad \epsilon = \frac{1}{2} [\nabla u + (\nabla u)^T] = \nabla^s u$$

$$\nabla \cdot ((\nu_1 \chi + (1 - \chi)\nu_2)\lambda_s; \epsilon) = 0 \qquad \epsilon = \epsilon_0 + \epsilon_f \text{ where } \epsilon_f = \nabla^s \phi$$

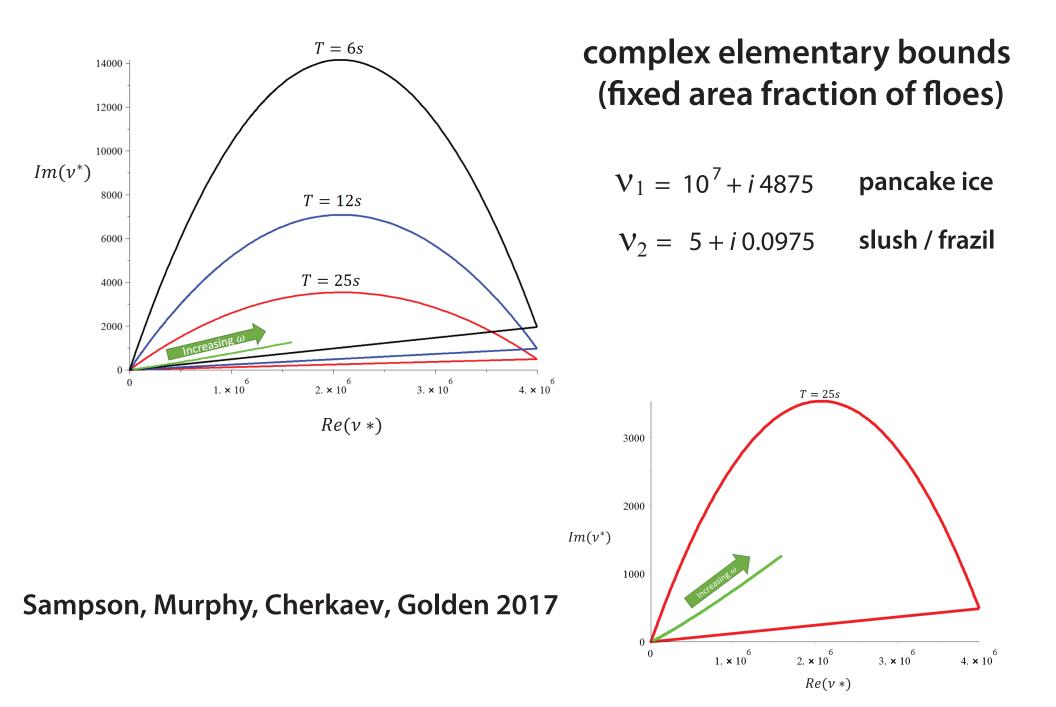
$$s = \frac{1}{1 - \frac{\nu_1}{\nu_2}} \qquad \text{Elasticity Tensor}$$

$$C_{ijkl}^* = \nu^* \left(\delta_{ik}\delta_{jl} + \delta_{il}\delta_{jk} - \frac{2}{3}\delta_{ij}\delta_{kl}\right) = \nu^*\lambda_s$$

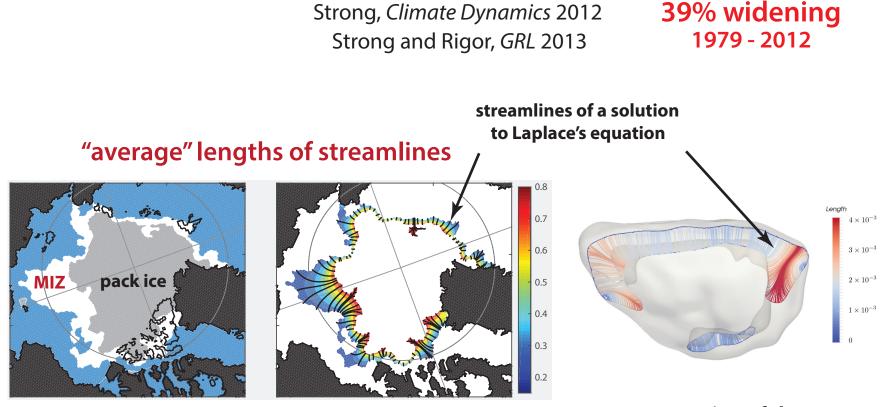
RESOLVENT
$$\epsilon = \left(1 - \frac{1}{s}\Gamma\chi\right)^{-1}\epsilon_0 \qquad \Gamma = \nabla^s (\nabla \cdot \nabla^s)^{-1} \nabla \cdot \epsilon_0 \text{ avg strain}$$

$$F(s) = 1 - \frac{\nu^*}{\nu_2} \qquad F(s) = ||\epsilon_0||^{-2} \int_{\Sigma} \frac{d\mu(\lambda)}{s - \lambda}$$

bounds on the effective complex viscoelasticity



Objective method for measuring MIZ width motivated by medical imaging and diagnostics



Arctic Marginal Ice Zone

crossection of the cerebral cortex of a rodent brain

analysis of different MIZ WIDTH definitions

Strong, Foster, Cherkaev, Eisenman, Golden J. Atmos. Oceanic Tech. 2017

> Strong and Golden Society for Industrial and Applied Mathematics News, April 2017

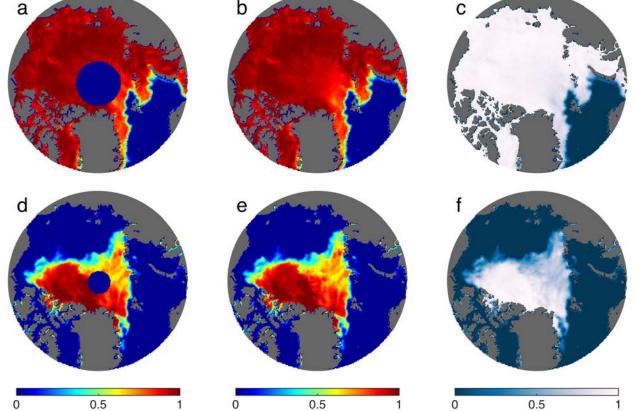
Filling the polar data gap

hole in satellite coverage of sea ice concentration field

previously assumed ice covered

Gap radius: 611 km 06 January 1985

Gap radius: 311 km 30 August 2007



fill with harmonic function satisfying satellite BC's plus stochastic term

Strong and Golden, *Remote Sensing* 2016 Strong and Golden, *SIAM News* 2017

Arctic and Antarctic field experiments

develop electromagnetic methods of monitoring fluid transport and microstructural transitions

extensive measurements of fluid and electrical transport properties of sea ice:

2007 Antarctic SIPEX	
2010 Antarctic McMu	urdo Sound
2011 Arctic Barro	w AK
2012 Arctic Barro	w AK
2012 Antarctic SIPEX	
2013 Arctic Barro	w AK
2014 Arctic Chuke	chi Sea



Notices Anterior Mathematical Society

of the American Mathematical Society

May 2009

Volume 56, Number 5

Climate Change and the Mathematics of Transport in Sea Ice

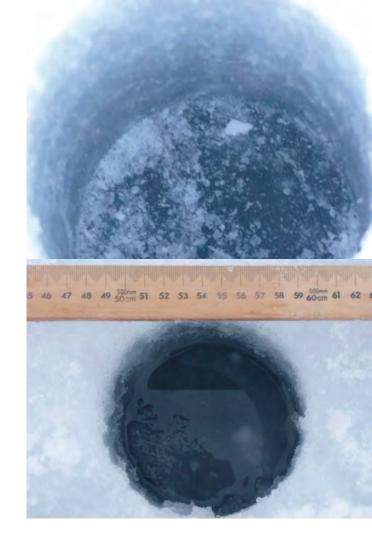
page 562

Mathematics and the Internet: A Source of Enormous Confusion and Great Potential

page 586

photo by Jan Lieser

Real analysis in polar coordinates (see page 613)



measuring fluid permeability of Antarctic sea ice

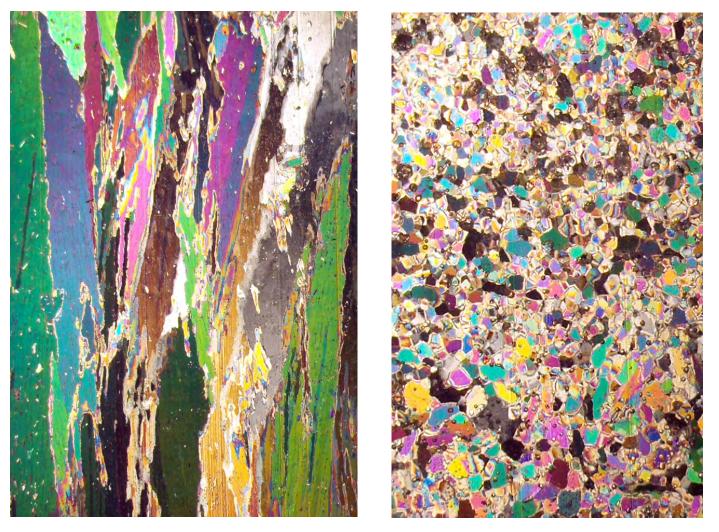
SIPEX 2007

higher threshold for fluid flow in Antarctic granular sea ice

columnar

5%

granular

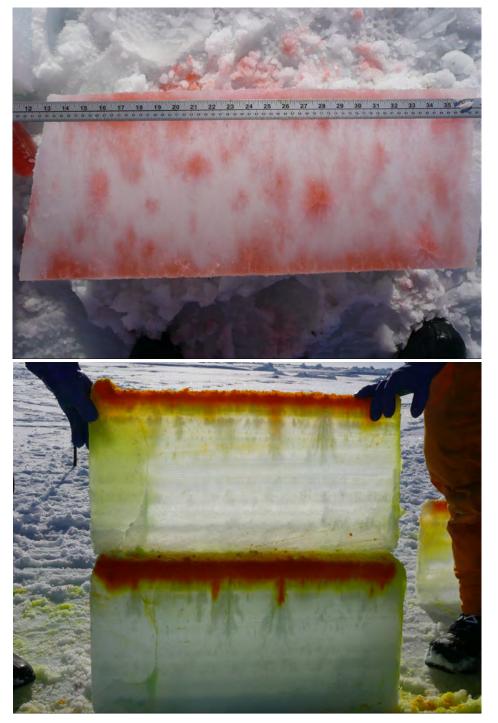


10%

Golden, Sampson, Gully, Lubbers, Tison 2017

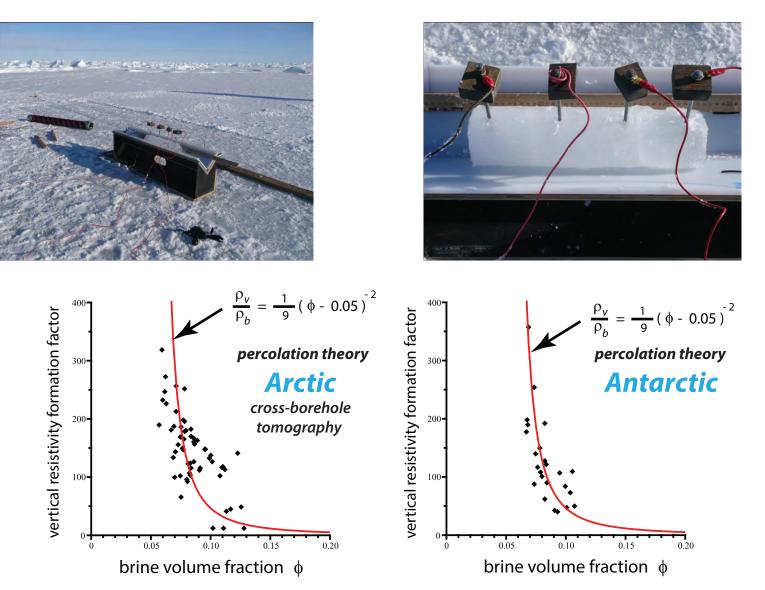
tracers flowing through inverted sea ice blocks







critical behavior of electrical transport in sea ice electrical signature of the on-off switch for fluid flow



cross-borehole tomography - electrical classification of sea ice layers

Golden, Eicken, Gully, Ingham, Jones, Lin, Reid, Sampson, Worby 2017

fractals and multiscale structure

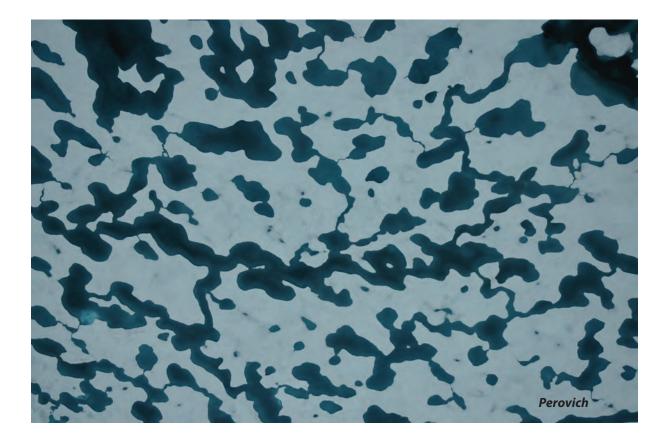


melt pond formation and albedo evolution:

- major drivers in polar climate
- key challenge for global climate models

numerical models of melt pond evolution, including topography, drainage (permeability), etc.

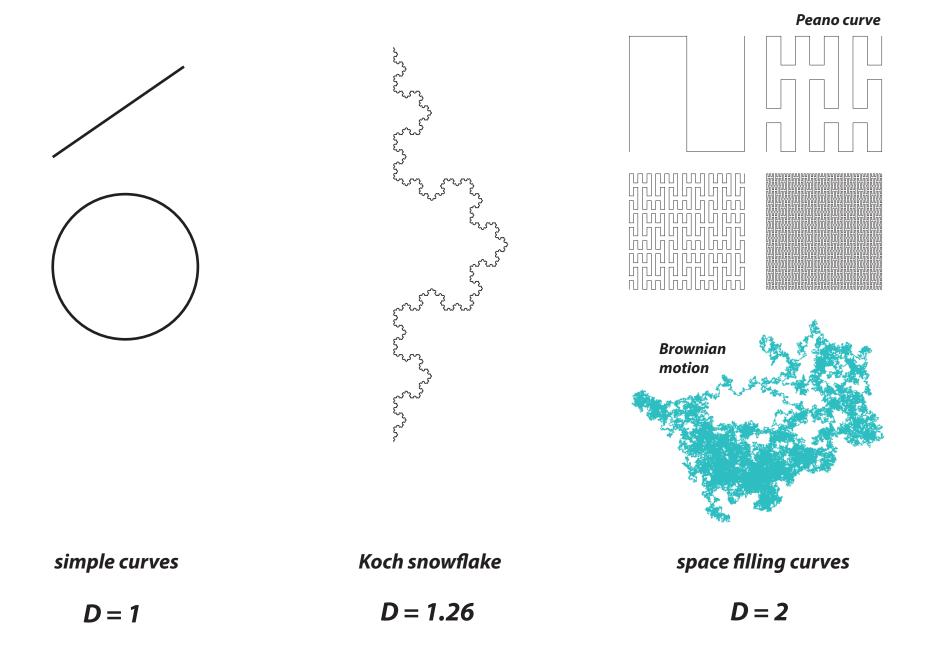
Lüthje, Feltham, Taylor, Worster 2006 Flocco, Feltham 2007 Skyllingstad, Paulson, Perovich 2009 Flocco, Feltham, Hunke 2012



Are there universal features of the evolution similar to phase transitions in statistical physics?

fractal curves in the plane

they wiggle so much that their dimension is >1



clouds exhibit fractal behavior from 1 to 1000 km



use *perimeter-area* data to find that cloud and rain boundaries are fractals

 $D \approx 1.35$

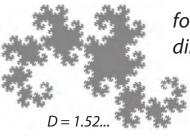
S. Lovejoy, Science, 1982

 $P \sim \sqrt{A}$

simple shapes

 $A = L^2$ $P = 4L = 4\sqrt{A}$

 $P \sim \sqrt{A}^{D}$



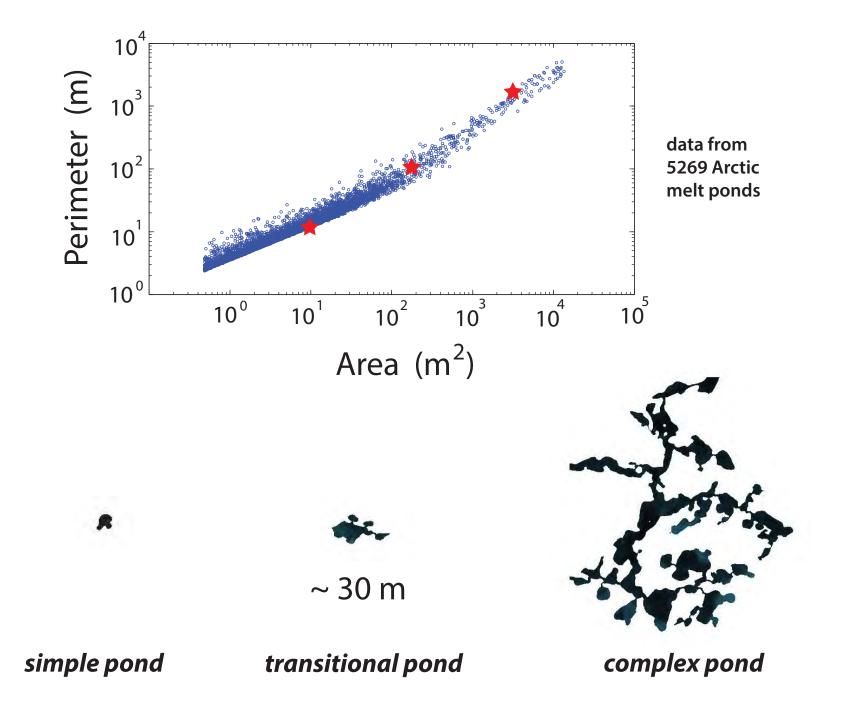
L

for fractals with dimension D

Transition in the fractal geometry of Arctic melt ponds

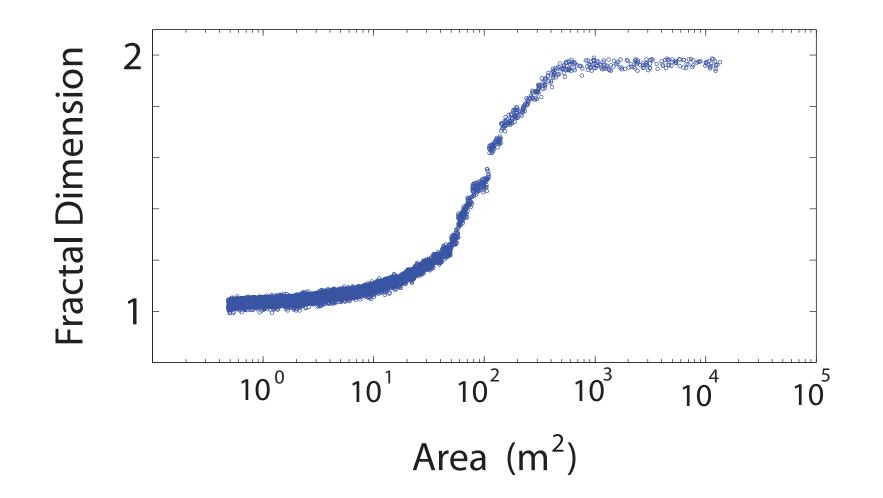
The Cryosphere, 2012

Christel Hohenegger, Bacim Alali, Kyle Steffen, Don Perovich, Ken Golden



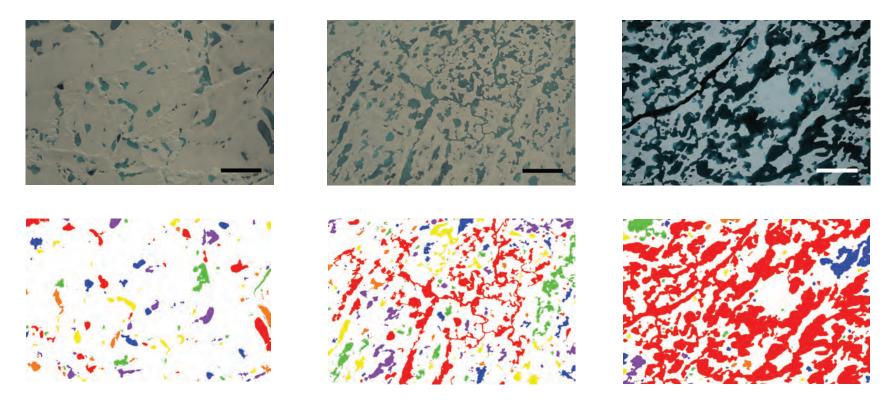
transition in the fractal dimension

complexity grows with length scale



compute "derivative" of area - perimeter data

small simple ponds coalesce to form large connected structures with complex boundaries



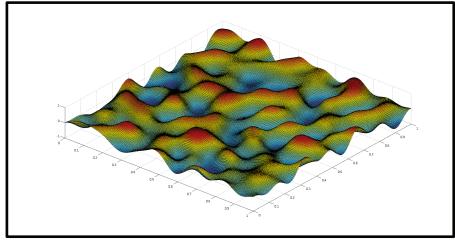
melt pond percolation

results on percolation threshold, correlation length, cluster behavior

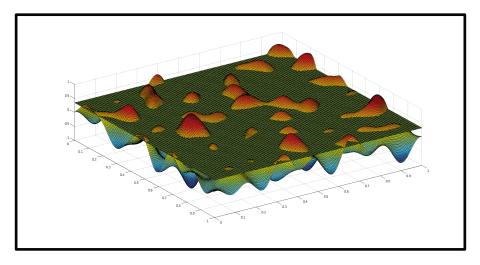
Anthony Cheng (Hillcrest HS), Dylan Webb (Skyline HS), Court Strong, Ken Golden

Continuum percolation model for melt pond evolution level sets of random surfaces

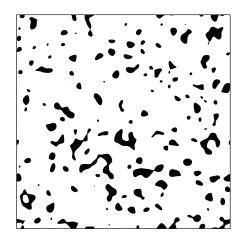
Brady Bowen, Court Strong, Ken Golden, J. Fractal Geometry 2017

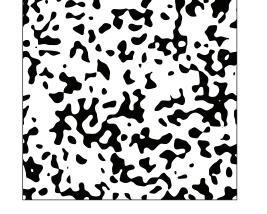


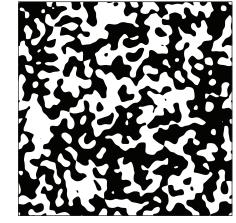
random Fourier series representation of surface topography



intersections of a plane with the surface define melt ponds







electronic transport in disordered media

diffusion in turbulent plasmas

Isichenko, Rev. Mod. Phys., 1992

melt pond evolution depends also on large-scale "pores" in ice cover

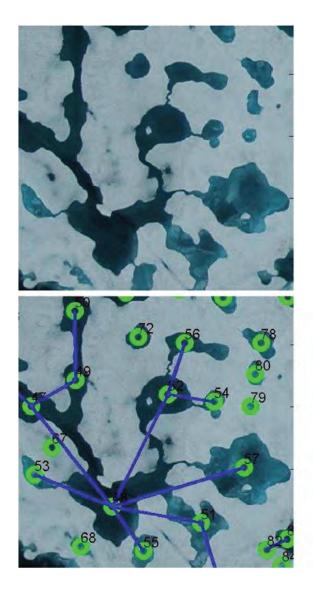
drainage vortex

photo courtesy of C. Polashenski and D. Perovich

Melt pond connectivity enables vast expanses of melt water to drain down seal holes, thaw holes, and leads in the ice.

Network modeling of Arctic melt ponds

Barjatia, Tasdizen, Song, Sampson, Golden Cold Regions Science and Tecnology, 2016



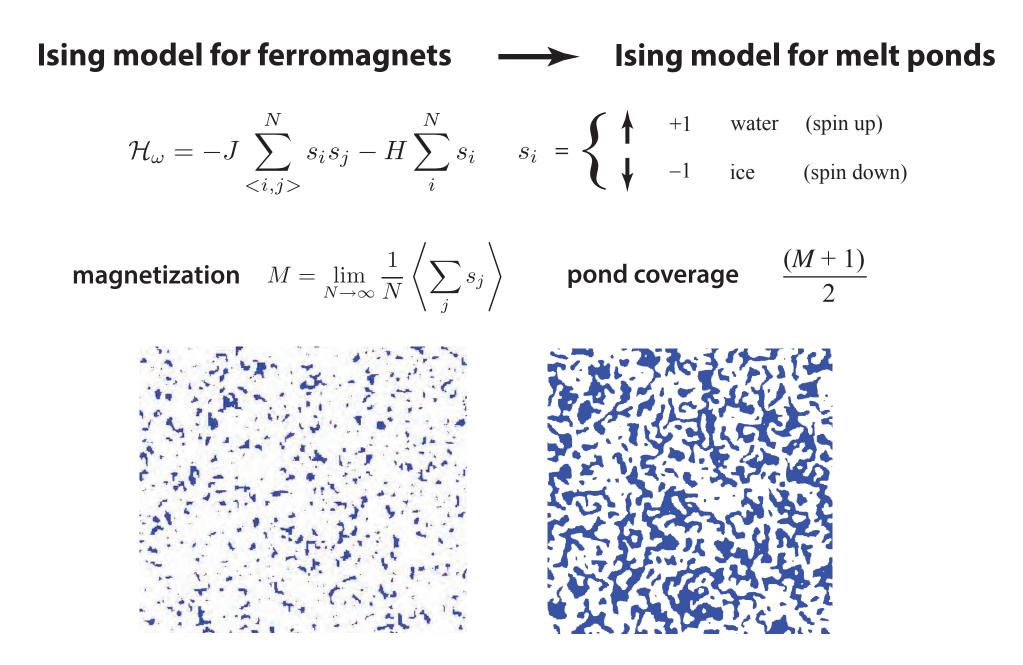
develop algorithms to map images of melt ponds onto

random resistor networks

graphs of nodes and edges with edge conductances

edge conductance ~ neck width

compute effective horizontal fluid conductivity



"melt ponds" are clusters of magnetic spins that align with the applied field

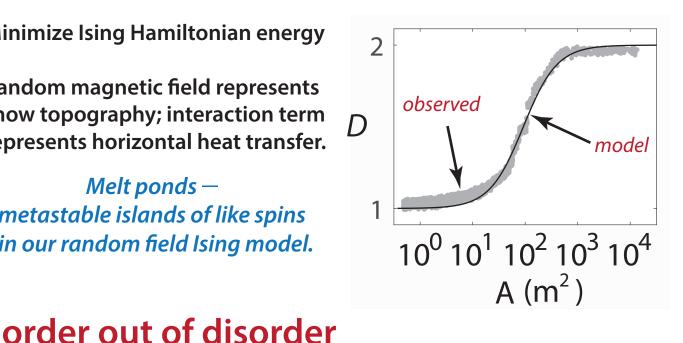
predictions of fractal transition, pond size exponent Ma, Sudakov, Strong, Golden 2017

Ising model results

Minimize Ising Hamiltonian energy

Random magnetic field represents snow topography; interaction term represents horizontal heat transfer.

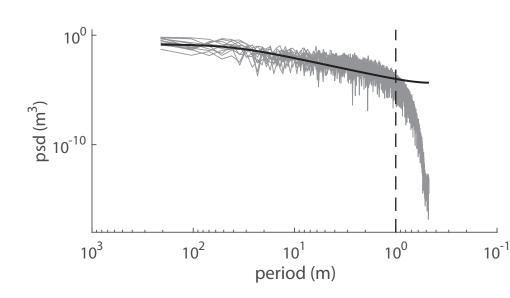
Melt ponds – metastable islands of like spins in our random field Ising model.



pond size distribution exponent

observed -1.5 (*Perovich*, *et al* 2002)

model -1.58



The lattice constant *a* must be small relative to the 10-20 m length scales prominent in sea ice and snow topography. We set a=1 m as the length scale above which important spatially correlated fluctuations occur in the power spectrum of snow topography.

The Melt Pond Conundrum:

How can ponds form on top of sea ice that is highly permeable?

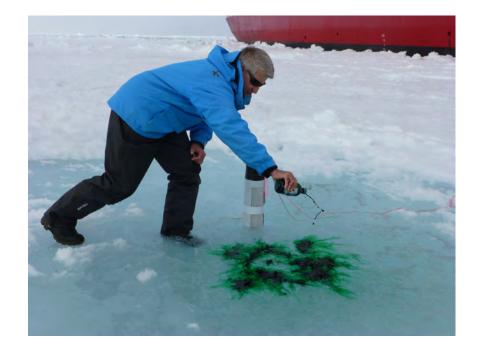
C. Polashenski, K. M. Golden, D. K. Perovich, E. Skyllingstad, A. Arnsten, C. Stwertka, N. Wright

Percolation Blockage: A Process that Enables Melt Pond Formation on First Year Arctic Sea Ice

J. Geophys. Res. Oceans 2017

2014 Study of Under Ice Blooms in the Chuckchi Ecosystem (SUBICE) aboard USCGC Healy





Conclusions

- 1. Summer Arctic sea ice is **melting rapidly**, and **melt ponds** and other processes must be accounted for in order to predict melting rates.
- 2. Fluid flow through sea ice mediates melt pond evolution and many processes important to climate change and polar ecosystems.
- 3. Homogenization and statistical physics help *link scales*, provide rigorous methods for finding effective behavior, and advance how sea ice is represented in climate models.
- 4. Critical behavior (in many forms) is inherent in the climate system.
- 5. Field experiments are essential to developing relevant mathematics.
- 6. Our research will help to improve projections of climate change, the fate of Earth's sea ice packs, and the ecosystems they support.

THANK YOU

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Arctic and Global Prediction Program Applied and Computational Analysis Program







Mathematics and Climate Research Network



Australian Government

Department of the Environment and Water Resources Australian Antarctic Division











Buchanan Bay, Antarctica Mertz Glacier Polynya Experiment July 1999