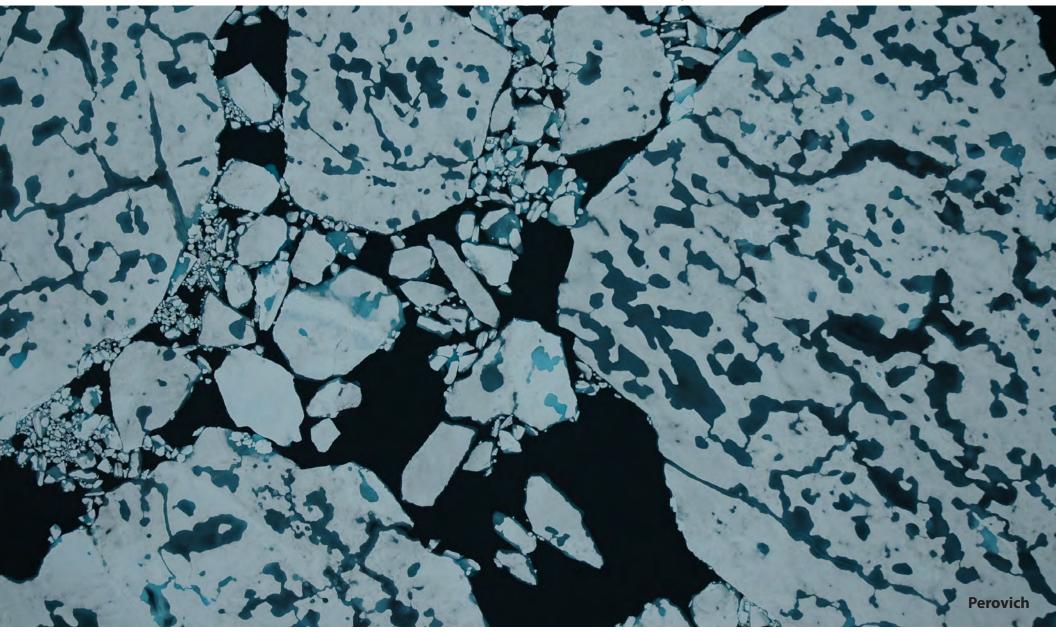
Multiscale homogenization for sea ice and other composite materials

Kenneth M. Golden University of Utah



ICIAM 2019, Valencia

SEA ICE covers ~12% of Earth's ocean surface

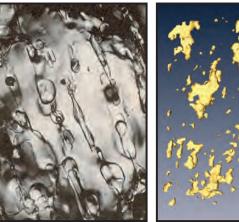
- boundary between ocean and atmosphere
- mediates exchange of heat, gases, momentum
- global ocean circulation
- hosts rich ecosystem
- indicator of climate change

polar ice caps critical to climate in reflecting sunlight during summer

Sea Ice is a Multiscale Composite Material

sea ice microstructure

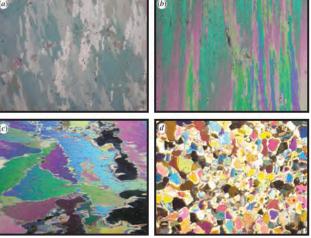
brine inclusions



Weeks & Assur 1969

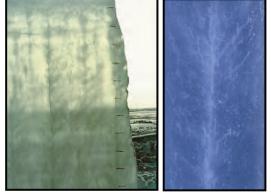
millimeters

polycrystals



Gully et al. Proc. Roy. Soc. A 2015

brine channels



D. Cole

K. Golden

sea ice mesostructure

H. Eicken

Golden et al. GRL 2007

sea ice macrostructure

centimeters

Arctic melt ponds



Antarctic pressure ridges

sea ice floes



sea ice pack



K. Frey

K. Golden

J. Weller



NASA

meters

What is this talk about? HOMOGENIZATION

What is the role of microstructure in determining effective properties?

Using methods of statistical physics and homogenization to LINK SCALES in the sea ice system ... rigorously compute effective behavior and improve climate models.

1. Sea ice microphysics and fluid transport

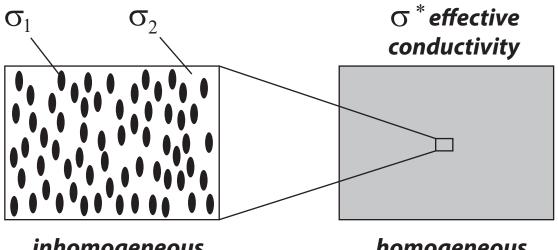
2. Analytic Continuation Method, integral representations

3. Extension of ACM to advection diffusion, waves in sea ice

4. Fractal geometry of melt pond evolution

Solving problems in physics of sea ice drives advances in theory of composite materials. cross - pollination

HOMOGENIZATION - Linking Scales in Composites



inhomogeneous medium homogeneous medium

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

Maxwell 1873 : effective conductivity of a dilute suspension of spheres Einstein 1906 : effective viscosity of a dilute suspension of rigid spheres in a fluid

Wiener 1912 : arithmetic and harmonic mean **bounds** on effective conductivity Hashin and Shtrikman 1962 : variational **bounds** on effective conductivity

widespread use of composites in late 20th century due in large part to advances in mathematically predicting their effective properties

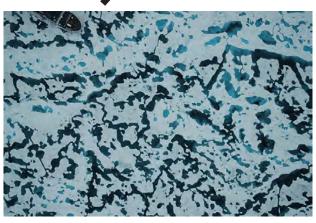
How do scales interact in the sea ice system?

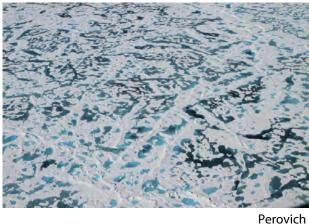


basin scale grid scale albedo

Linking Scales

km scale melt ponds





Scales

km scale melt ponds

Linking

mm scale brine inclusions





meter scale snow topography

sea ice microphysics

fluid transport

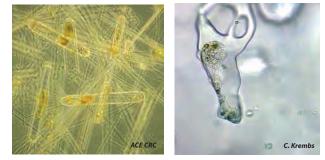
fluid flow through the porous microstructure of sea ice governs key processes in polar climate and ecosystems

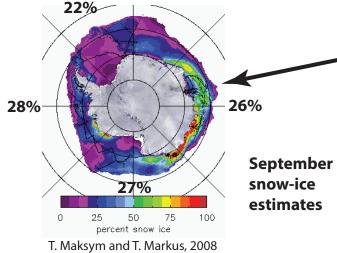
evolution of Arctic melt ponds and sea ice albedo



nutrient flux for algal communities







Antarctic surface flooding and snow-ice formation

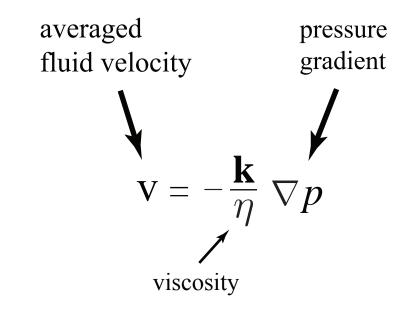
evolution of salinity profiles
ocean-ice-air exchanges of heat, CO₂

fluid permeability of a porous medium



Darcy's Law

for slow viscous flow in a porous medium



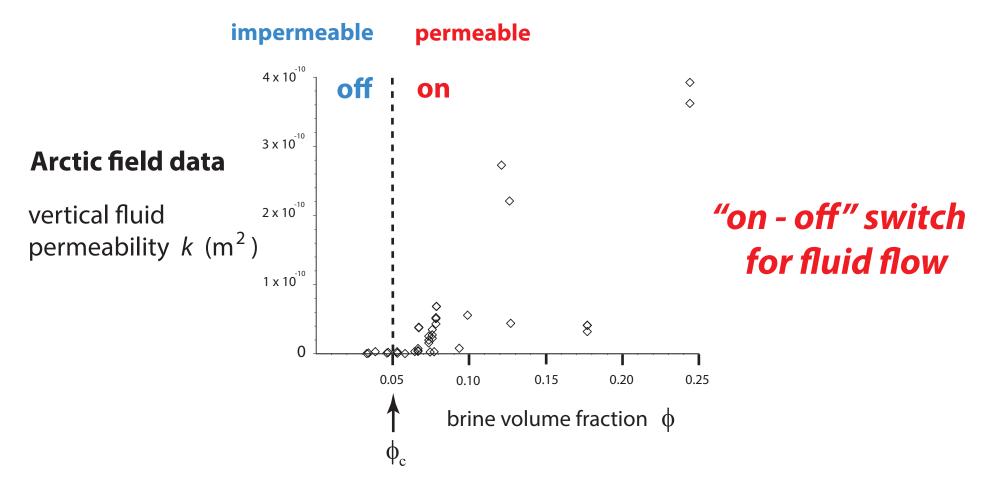
how much water gets through the sample per unit time?

k = fluid permeability tensor

HOMOGENIZATION

mathematics for analyzing effective behavior of heterogeneous systems

Critical behavior of fluid transport in sea ice



critical brine volume fraction $\phi_c \approx 5\%$ \checkmark $T_c \approx -5^{\circ}C, S \approx 5$ ppt

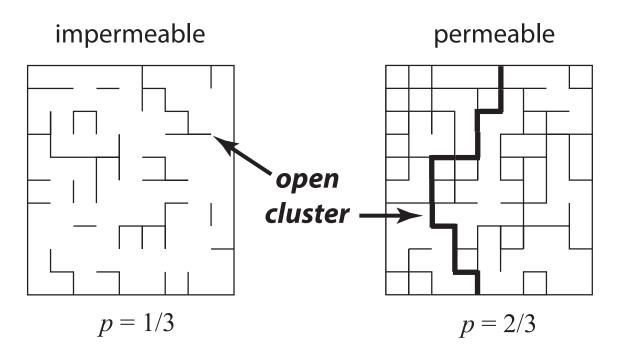
RULE OF FIVES

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

Why is the rule of fives true?

percolation theory

probabilistic theory of connectedness



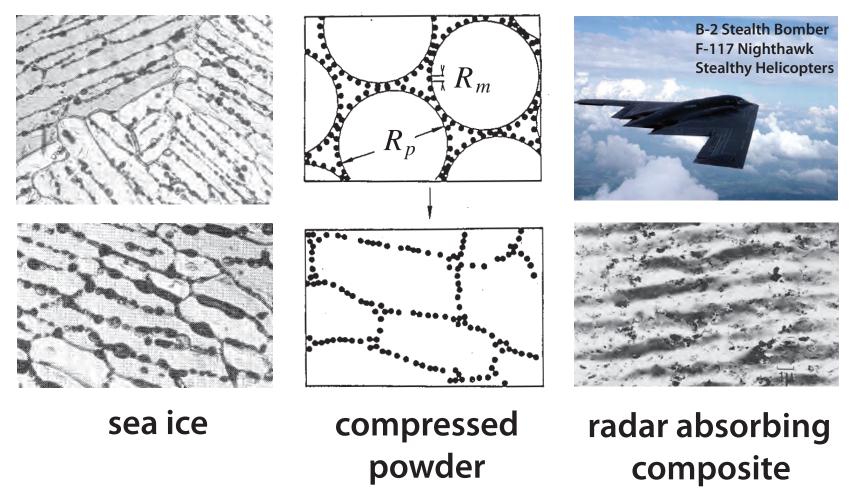
bond \longrightarrow *open with probability p closed with probability 1-p*

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

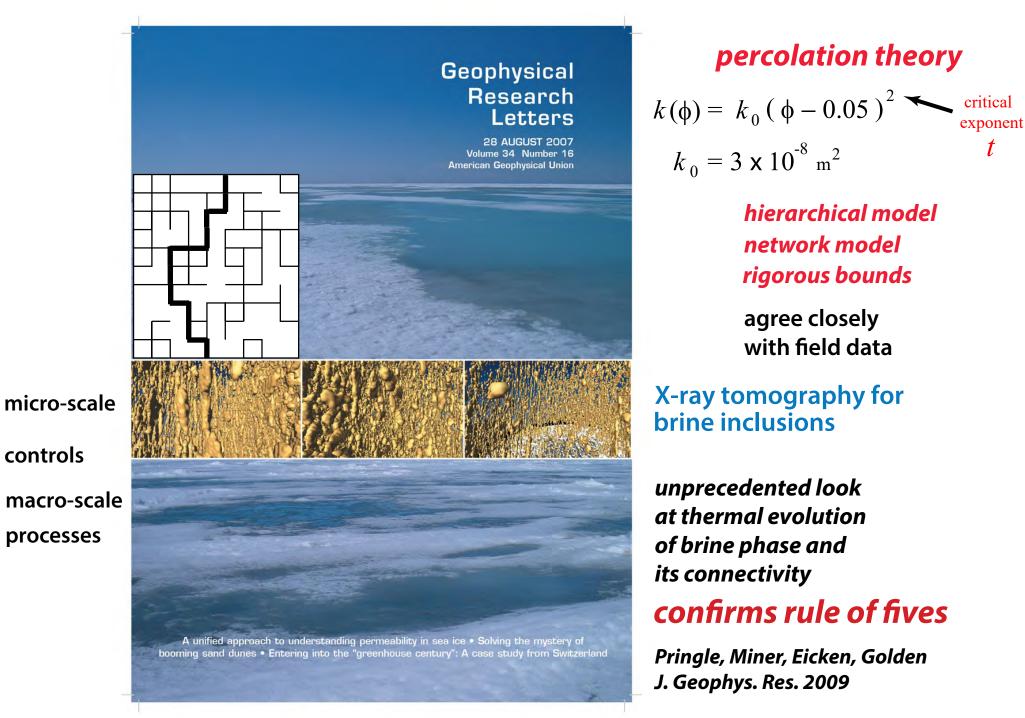
 $\phi_c \approx 5\%$ Golden, Ackley, Lytle, *Science*, 1998



sea ice is radar absorbing

Thermal evolution of permeability and microstructure in sea ice

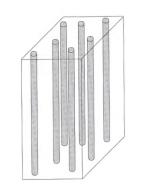
Golden, Eicken, Heaton, Miner, Pringle, Zhu, Geophysical Research Letters 2007

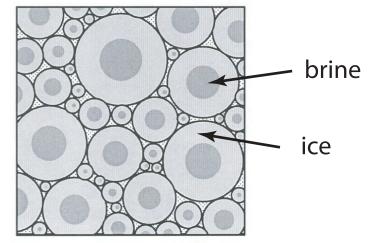


PIPE BOUNDS on vertical fluid permeability k

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

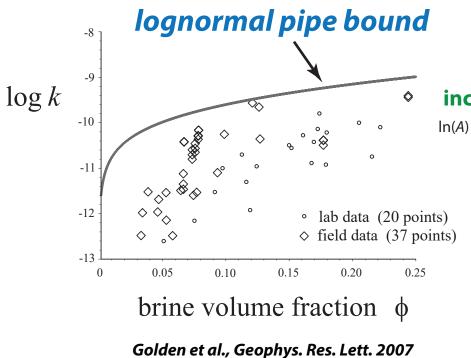
> vertical pipes with appropriate radii maximize k





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

optimal coated cylinder geometry



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

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

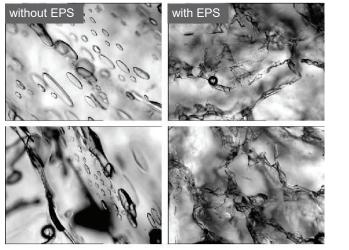
 $\mathbf{k} \leq \gamma^{-1} \mathbf{I}$

for any ergodic porous medium (Torquato 2002, 2004)

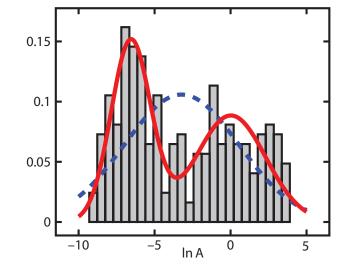
BACTERIAL FORAGING

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

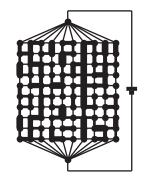
How does EPS affect fluid transport?



Krembs, Eicken, Deming, PNAS 2011



RANDOM PIPE MODEL



 $R_{i,j}^{h} \xrightarrow{R_{i,j}^{v}} R_{i,j}^{h}$

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, 2018

How does the biology affect the physics?

Notices Notes Series

of the American Mathematical Society

May 2009

Volume 56, Number 5

Climate Change and the Mathematics of Transport in Sea Ice

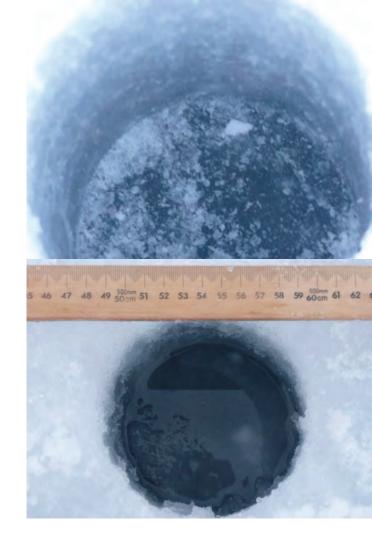
page 562

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)

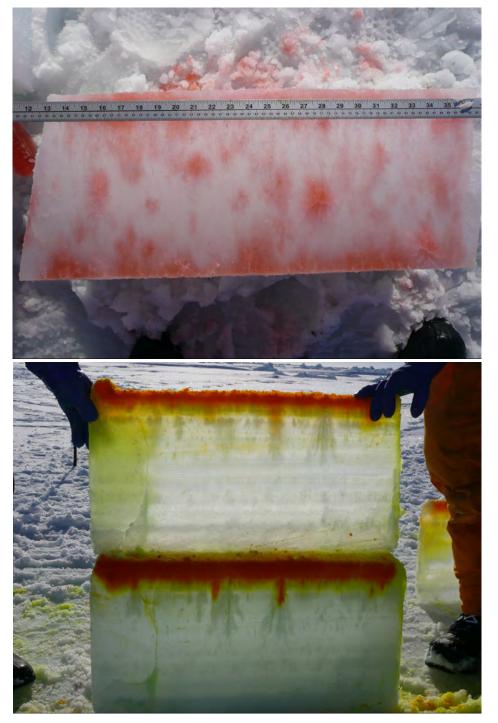


measuring fluid permeability of Antarctic sea ice

SIPEX 2007

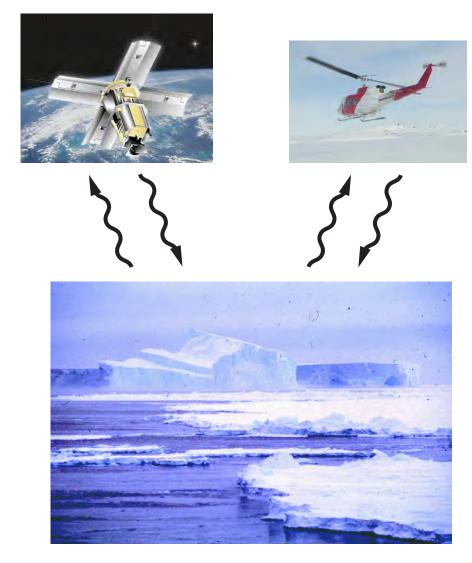
tracers flowing through inverted sea ice blocks







Remote sensing of sea ice



sea ice thickness ice concentration

INVERSE PROBLEM

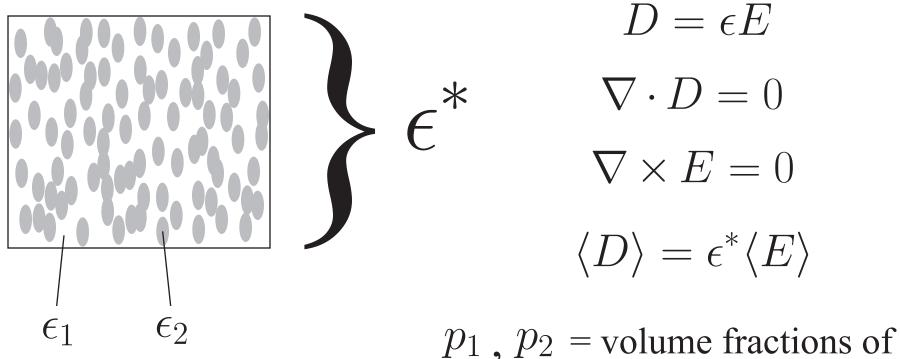
Recover sea ice properties from electromagnetic (EM) data

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



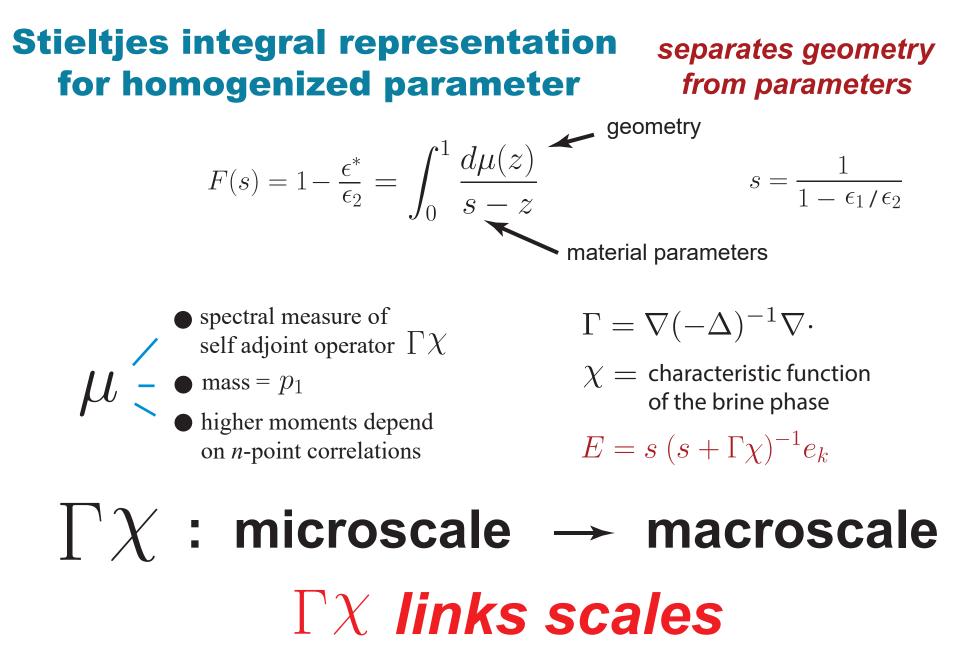
the components

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

What are the effective propagation characteristics of an EM wave (radar, microwaves) in the medium?

Analytic Continuation Method for Homogenization

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

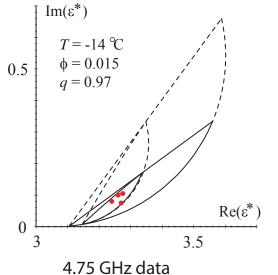


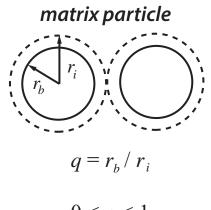
Golden and Papanicolaou, Comm. Math. Phys. 1983

forward and inverse bounds on the complex permittivity of sea ice







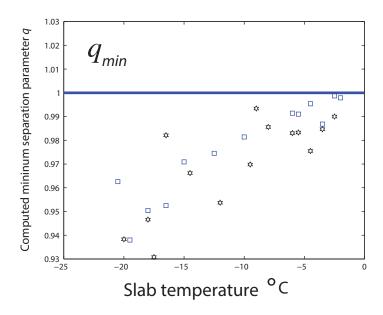


0 < q < 1

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*

direct calculation of spectral measures

Murphy, Hohenegger, Cherkaev, Golden, Comm. Math. Sci. 2015

- depends only on the composite geometry
- discretization of microstructural image gives binary network
- fundamental operator becomes a random matrix
- spectral measure computed from eigenvalues and eigenvectors

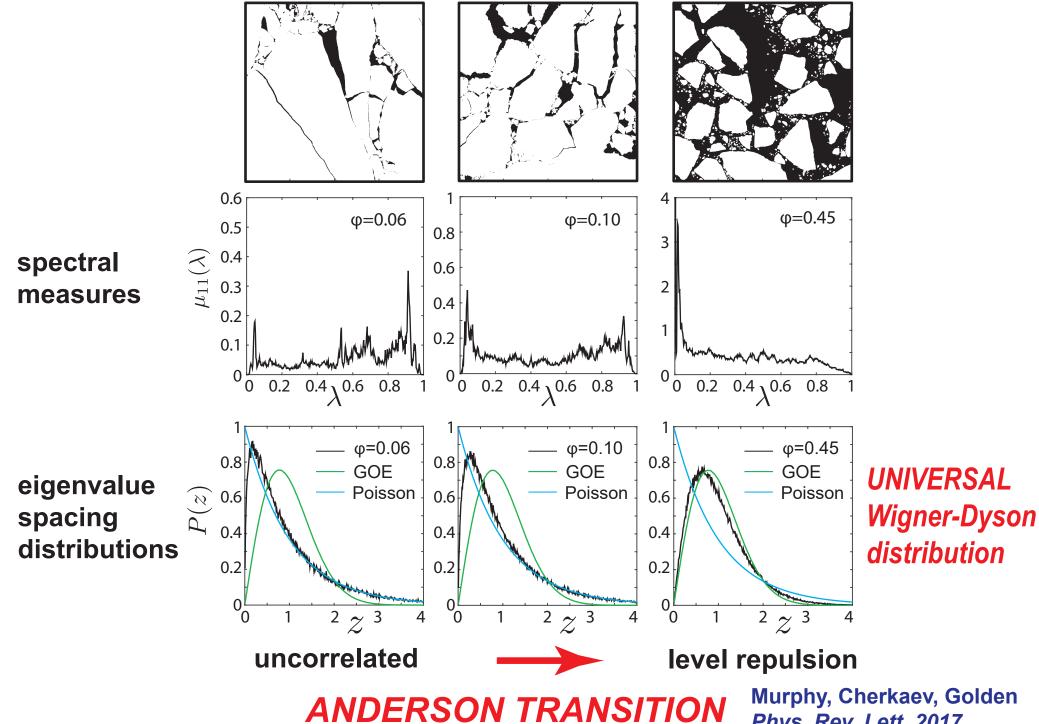
once we have the spectral measure μ it can be used in Stieltjes integrals for other transport coefficients:

electrical and thermal conductivity, complex permittivity, magnetic permeability, diffusion, fluid flow properties

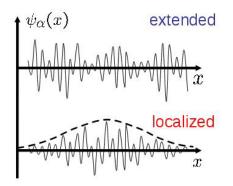
earlier studies of spectral measures

Day and Thorpe 1996 Helsing, McPhedran, Milton 2011

Spectral computations for sea ice floe configurations



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



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 ! --

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 orientation statistics
- 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

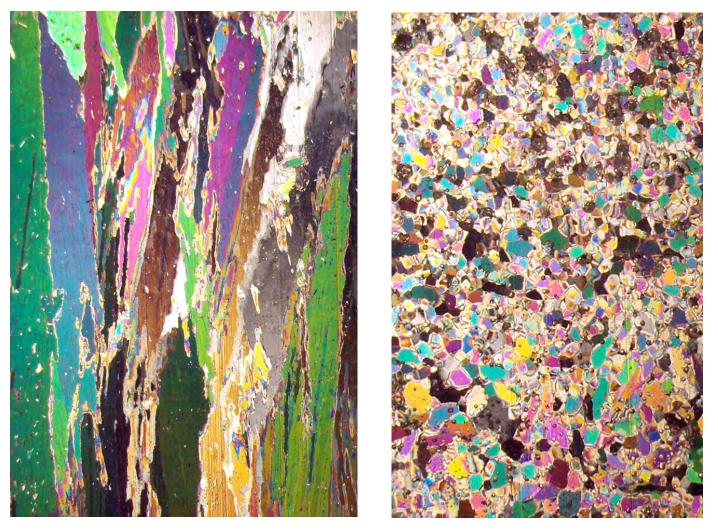


higher threshold for fluid flow in Antarctic granular sea ice

columnar

5%

granular



10%

Golden, Sampson, Gully, Lubbers, Tison 2019

advection enhanced diffusion

effective diffusivity

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

advection diffusion equation with a velocity field $\,\vec{u}\,$

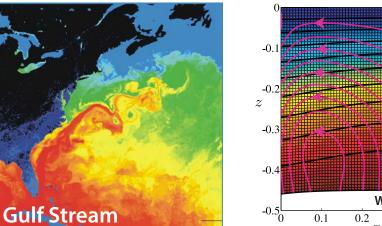
$$\frac{\partial T}{\partial t} + \vec{u} \cdot \vec{\nabla} T = \kappa_0 \Delta T$$
$$\vec{\nabla} \cdot \vec{u} = 0$$
$$homogenize$$
$$\frac{\partial \overline{T}}{\partial t} = \kappa^* \Delta \overline{T}$$

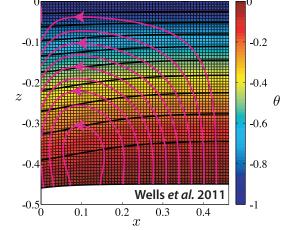
κ^{*} effective diffusivity

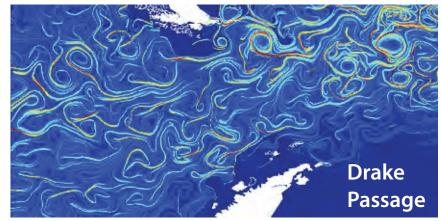
Stieltjes integral for κ^* with spectral measure

Avellaneda and Majda, PRL 89, CMP 91

Murphy, Cherkaev, Xin, Zhu, Golden, *Ann. Math. Sci. Appl.* 2017 Murphy, Cherkaev, Zhu, Xin, Golden, *J. Math. Phys.* 2019









Stieltjes Integral Representation for Advection Diffusion

Murphy, Cherkaev, Zhu, Xin, Golden, J. Math. Phys. 2019

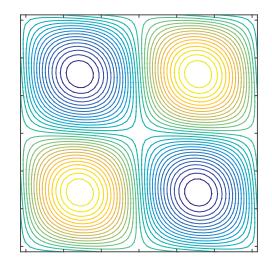
$$\kappa^* = \kappa \left(1 + \int_{-\infty}^{\infty} \frac{d\mu(\tau)}{\kappa^2 + \tau^2} \right), \quad F(\kappa) = \int_{-\infty}^{\infty} \frac{d\mu(\tau)}{\kappa^2 + \tau^2}$$

- μ is a positive definite measure corresponding to the spectral resolution of the self-adjoint operator $i\Gamma H\Gamma$
- H = stream matrix , $\kappa =$ local diffusivity
- $\Gamma :=
 abla (-\Delta)^{-1}
 abla \cdot$, Δ is the Laplace operator
- $i\Gamma H\Gamma$ is bounded for time independent flows
- $F(\kappa)$ is analytic off the spectral interval in the κ -plane

separation of material properties and flow field spectral measure calculations

Rigorous bounds on convection enhanced thermal conductivity of sea ice

Kraitzman, Hardenbrook, Murphy, Zhu, Cherkaev, Strong, Golden 2019

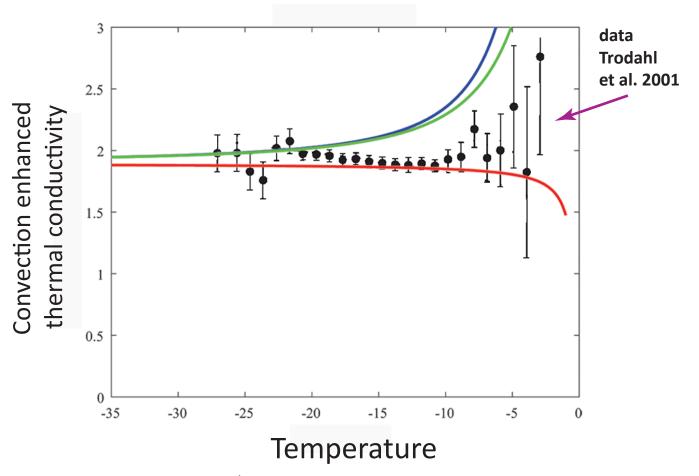


cat's eye flow model for brine convection cells

similar bounds for shear flows

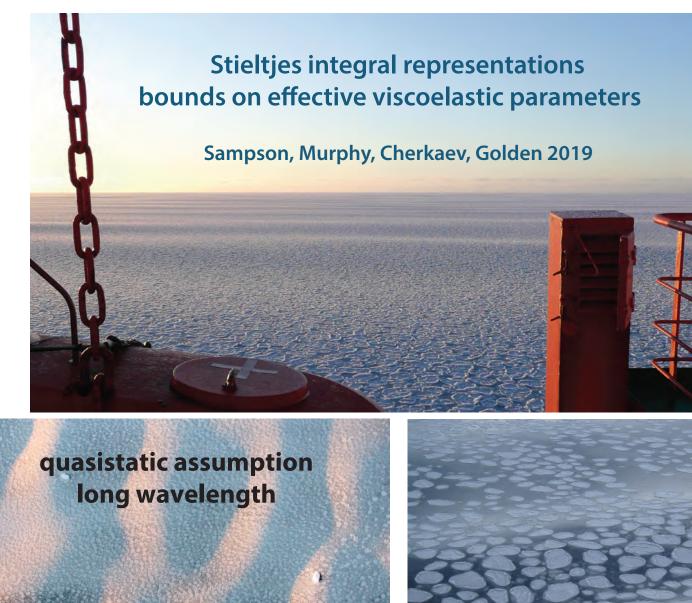
rigorous bounds assuming information on flow field INSIDE inclusions

Kraitzman, Cherkaev, Golden SIAM J. Appl. Math (in revision), 2019



rigorous Pade bounds from Stieltjes integral + analytical calculations of moments of measure

wave propagation in the marginal ice zone



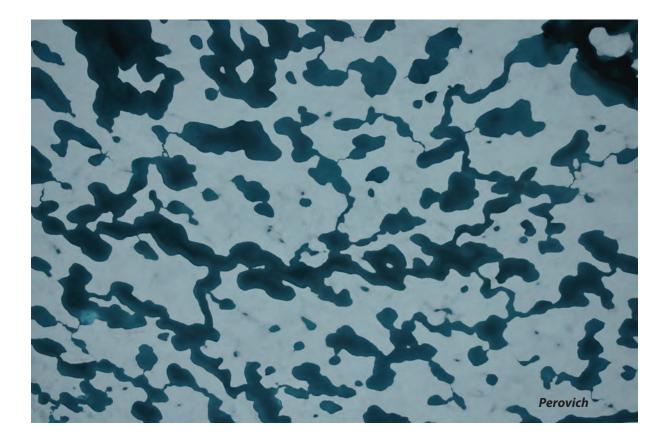


melt pond formation and albedo evolution:

- 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

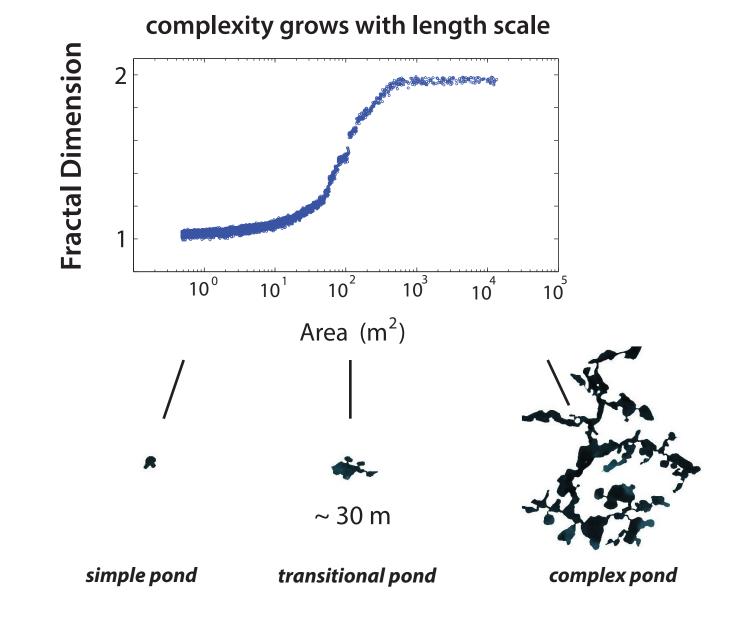


Are there universal features of the evolution similar to phase transitions in statistical physics?

Transition in the fractal geometry of Arctic melt ponds

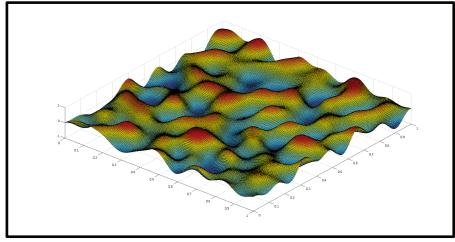
Christel Hohenegger, Bacim Alali, Kyle Steffen, Don Perovich, Ken Golden

The Cryosphere, 2012

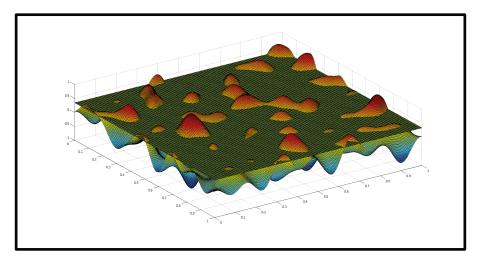


Continuum percolation model for melt pond evolution level sets of random surfaces

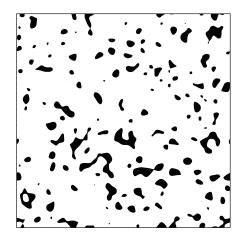
Brady Bowen, Court Strong, Ken Golden, J. Fractal Geometry 2018

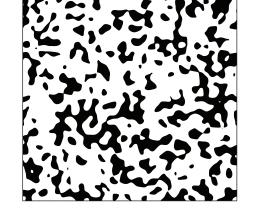


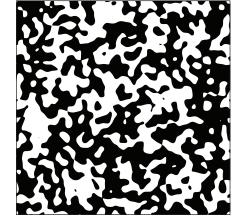
random Fourier series representation of surface topography



intersections of a plane with the surface define melt ponds





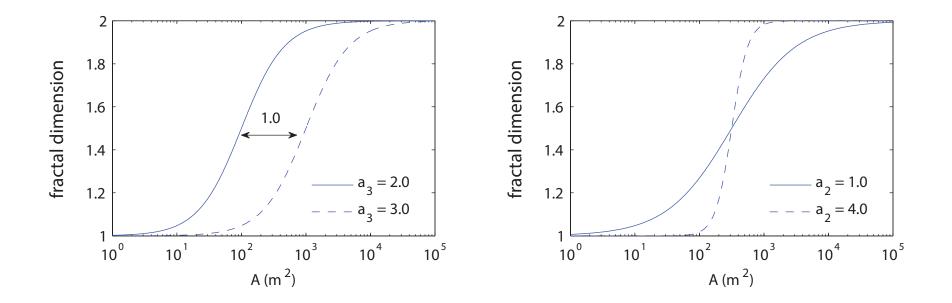


electronic transport in disordered media

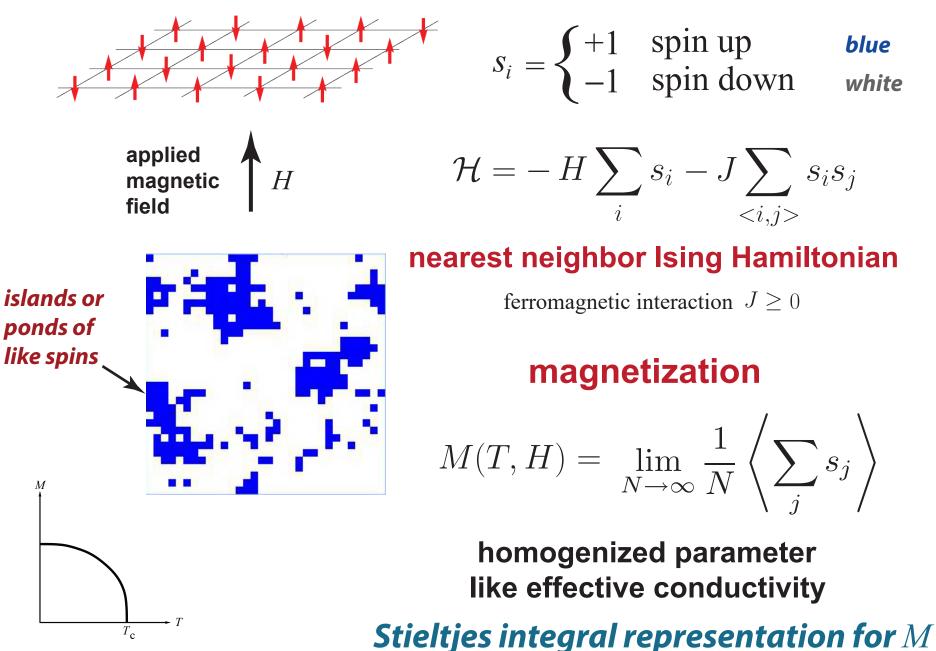
diffusion in turbulent plasmas

Isichenko, Rev. Mod. Phys., 1992

fractal dimension curves depend on statistical parameters defining random surface

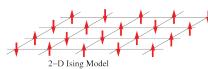


Ising Model for a Ferromagnet



Curie point critical temperature

Baker, PRL 1968



Ising model for ferromagnets —> Ising model for melt ponds

Ma, Sudakov, Strong, Golden, New J. Phys. 2019

 $\mathcal{H}_{\omega} = -J \sum_{\langle i,j \rangle}^{N} s_i s_j - \sum_{i}^{N} H_i s_i \qquad s_i = \begin{cases} \bigstar & +1 & \text{water (spin up)} \\ \checkmark & -1 & \text{ice (spin down)} \end{cases}$

magnetization $M = \lim_{N \to \infty} \frac{1}{N} \left\langle \sum_{i} s_{i} \right\rangle$ pond coverage $\frac{(M+1)}{2}$

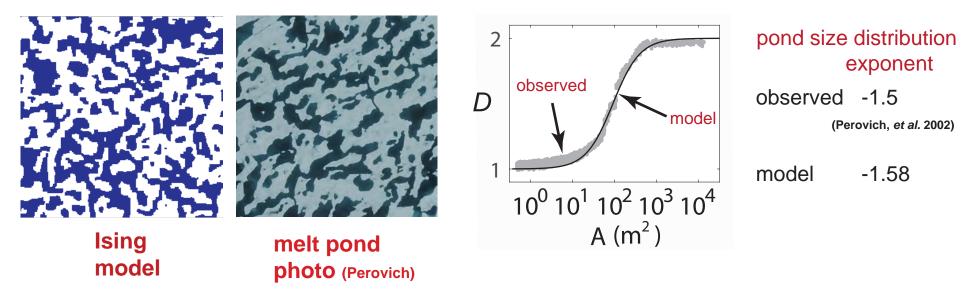
random magnetic field represents snow topography

only nearest neighbor patches interact

Starting with random initial configurations, as Hamiltonian energy is minimized by Glauber spin flip dynamics, system "flows" toward metastable equilibria.

Melt ponds are metastable islands of like spins.

Order from Disorder



ONLY MEASURED INPUT = LENGTH SCALE (GRID SIZE) from snow topography data

The distribution of solar energy under ponded first-year sea ice

Horvat, Flocco, Rees Jones, Roach, Golden, in revision, 2019

- Model for 3D light field under ponded sea ice.
- Distribution of solar energy at depth influenced by *shape and connectivity* of melt ponds, as well as area fraction.
- Aggregate properties of the sub-ice light field, such as a significant enhancement of available solar energy under the ice, are controlled by parameter closely related to pond fractal geometry.
- Model and analysis explain how melt pond geometry *homogenizes* under-ice light field, affecting habitability.

Pond geometry affects the ecology of the Arctic Ocean.

The Melt Pond Conundrum:

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

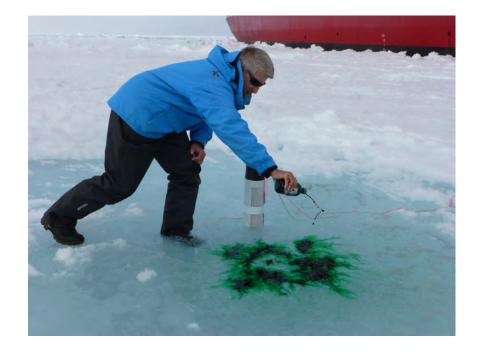
C. Polashenski, K. M. Golden, D. K. Perovich, E. Skyllingstad, A. Arnsten, C. Stwertka, N. Wright

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. Sea ice is a fascinating multiscale composite with structure similar to many other natural and man-made materials.
- 2. Mathematical methods developed for sea ice advance the theory of composites in general.
- 2. Homogenization and statistical physics help *link scales in sea ice and composites*; provide rigorous methods for finding effective behavior; advance sea ice representations in climate models.
- 3. Fluid flow through sea ice mediates melt pond evolution and many processes important to climate change and polar ecosystems.
- 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

Office of Naval Research

Applied and Computational Analysis Program Arctic and Global Prediction Program

National Science Foundation

Division of Mathematical Sciences Division of Polar Programs











Australian Government

Department of the Environment and Water Resources Australian Antarctic Division





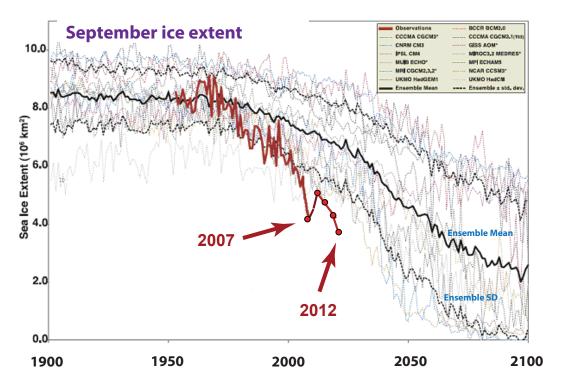






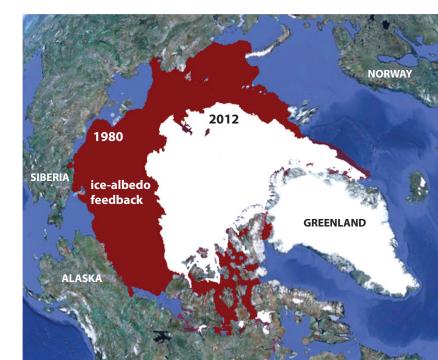
Buchanan Bay, Antarctica Mertz Glacier Polynya Experiment July 1999

Arctic sea ice decline: faster than predicted by climate models



Change in Arctic Sea Ice Extent

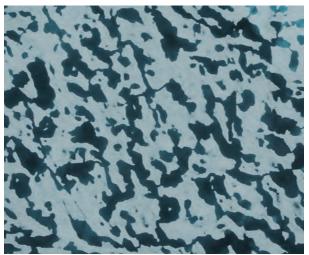
September 1980 -- 7.8 million square kilometers September 2012 -- 3.4 million square kilometers Stroeve et al., GRL, 2007 Stroeve et al., GRL, 2012



challenge

represent sea ice more realistically in climate models account for key processes such as melt pond evolution

How do patterns of dark and light evolve?



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*