Homogenization, Anderson Transitions, and Waves in Sea Ice

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sea ice is a multiscale composite

structured on many length scales - from tenths of mm's to tens of km's



millimeters



centimeters

pancakes







ice floes

meters

kilometers

What is this talk about? HOMOGENIZATION

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

Find unexpected Anderson transition in composites along the way!

1. Sea ice microphysics and fluid transport homogenization and percolation theory

- 2. EM monitoring of sea ice, analytic continuation method random matrix theory and Anderson transitions
- 3. Extension of ACM to advection diffusion, waves in sea ice Stieltjes integral representations, spectral measures

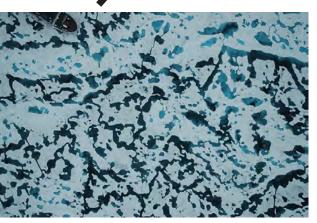
How do scales interact in the sea ice system?



basin scale grid scale albedo

Linking Scales

km scale melt ponds





km scale melt ponds

Linking

mm scale brine inclusions



Scales



meter scale snow topography

sea ice microphysics

fluid transport

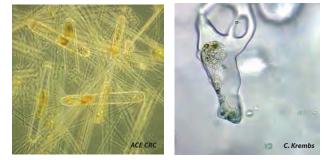
fluid flow through the porous microstructure of sea ice governs key processes in polar climate and ecosystems

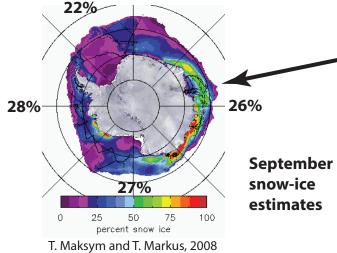
evolution of Arctic melt ponds and sea ice albedo



nutrient flux for algal communities



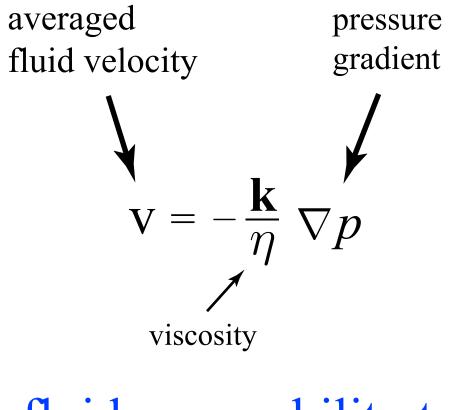




Antarctic surface flooding and snow-ice formation

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

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

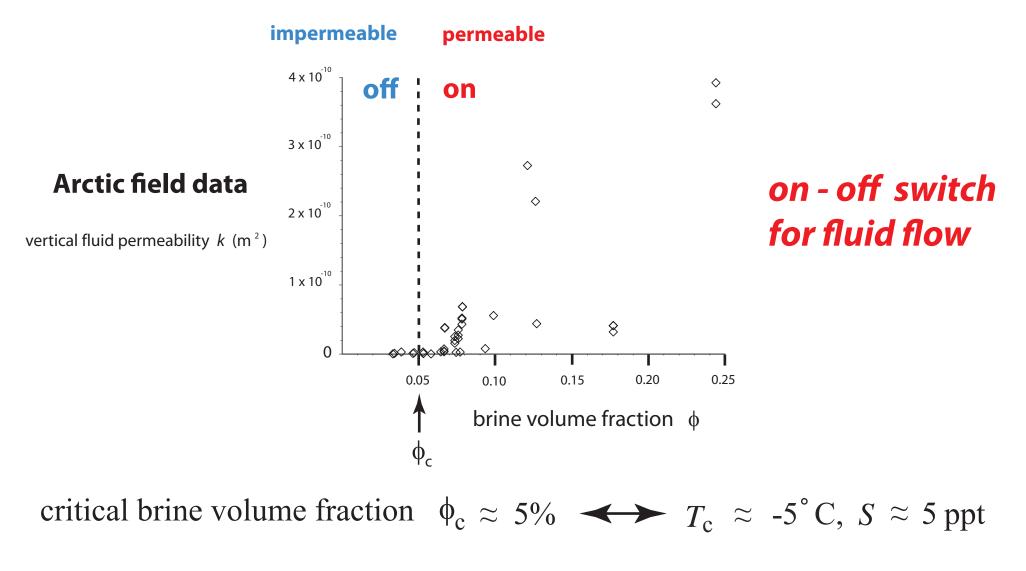


k = fluid permeability tensor example of *homogenization*

mathematics for analyzing effective behavior of heterogeneous systems

e.g. transport properties of composites - electrical conductivity, thermal conductivity, etc.

Critical behavior of fluid transport in sea ice

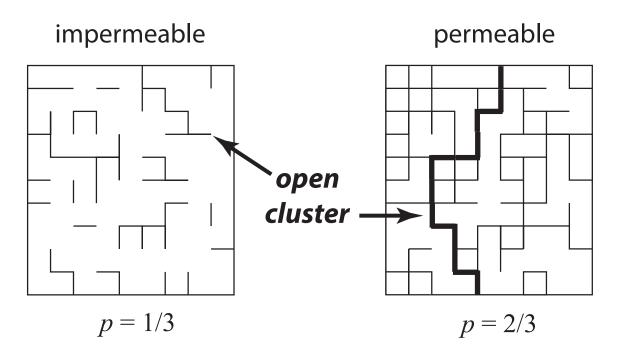


RULE OF FIVES

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

percolation theory

probabilistic theory of connectedness



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

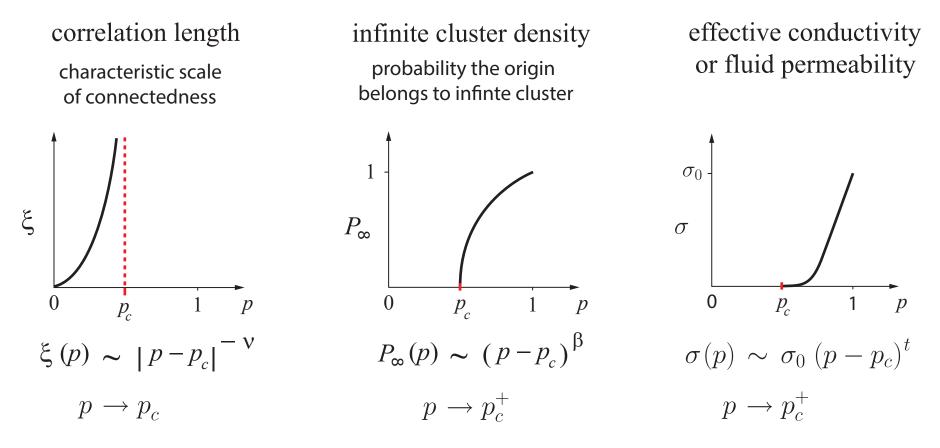
percolation threshold $p_c = 1/2$ for d = 2

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

order parameters in percolation theory

geometry

transport



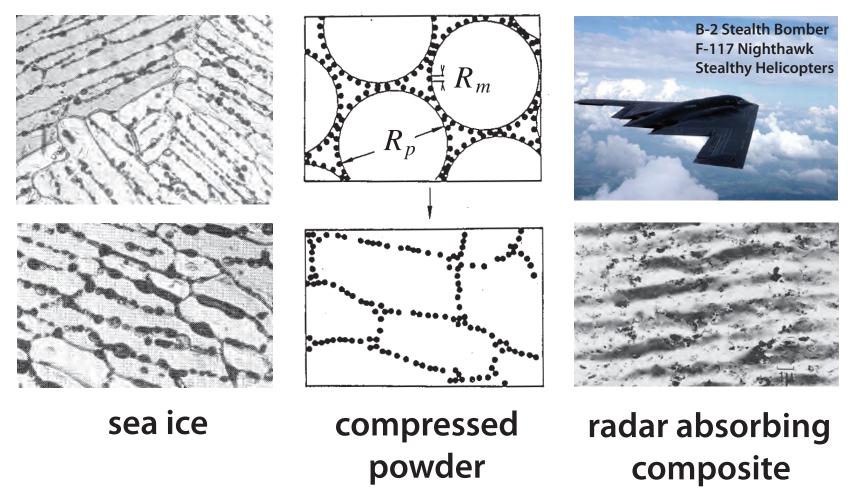
UNIVERSAL critical exponents for lattices -- depend only on dimension

 $1 \le t \le 2$ (for idealized model), Golden, *Phys. Rev. Lett.* 1990; *Comm. Math. Phys.* 1992

non-universal behavior in continuum

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

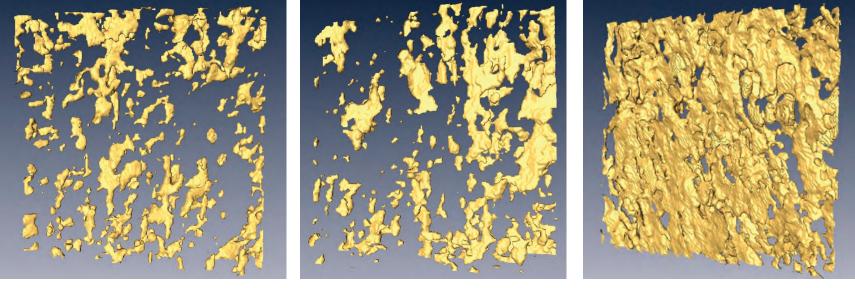
controls

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

Remote sensing of sea ice



sea ice thickness ice concentration

INVERSE PROBLEM

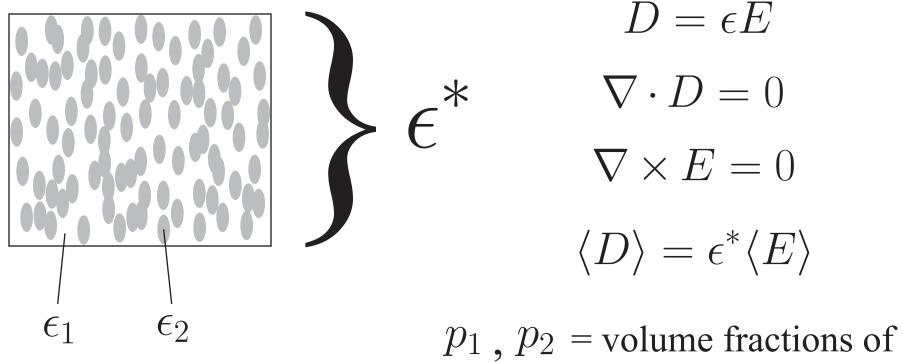
Recover sea ice properties from electromagnetic (EM) data

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

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

Herglotz function

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

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

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

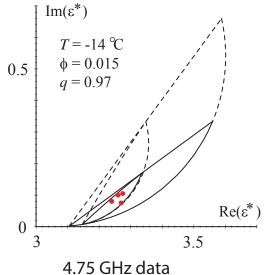
$$E = (s + \Gamma \chi)^{-1} e_k$$

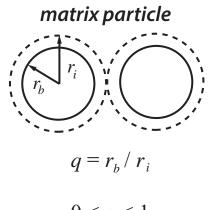
$\Gamma \chi$: microscale \rightarrow macroscale $\Gamma \chi$ *links scales*

forward and inverse bounds on the complex permittivity of sea ice







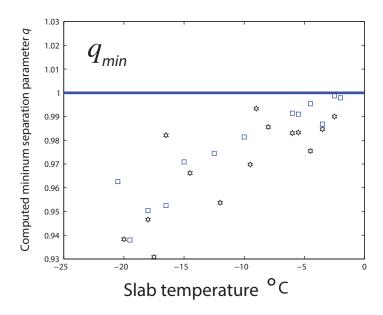


0 < q < 1

Golden 1995, 1997 Bruno 1991

inverse bounds and recovery of brine porosity

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



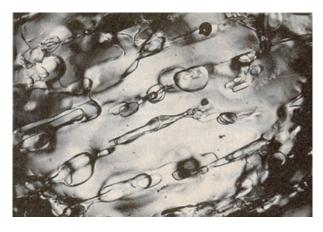
inversion for brine inclusion separations in sea ice from measurements of effective complex permittivity ϵ^*

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

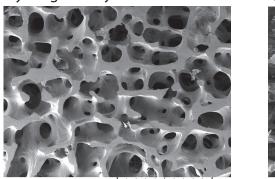


young healthy trabecular bone



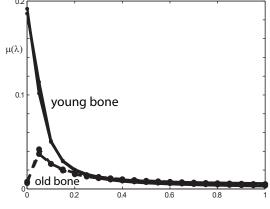
spectral characterization of porous microstructures in human bone

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

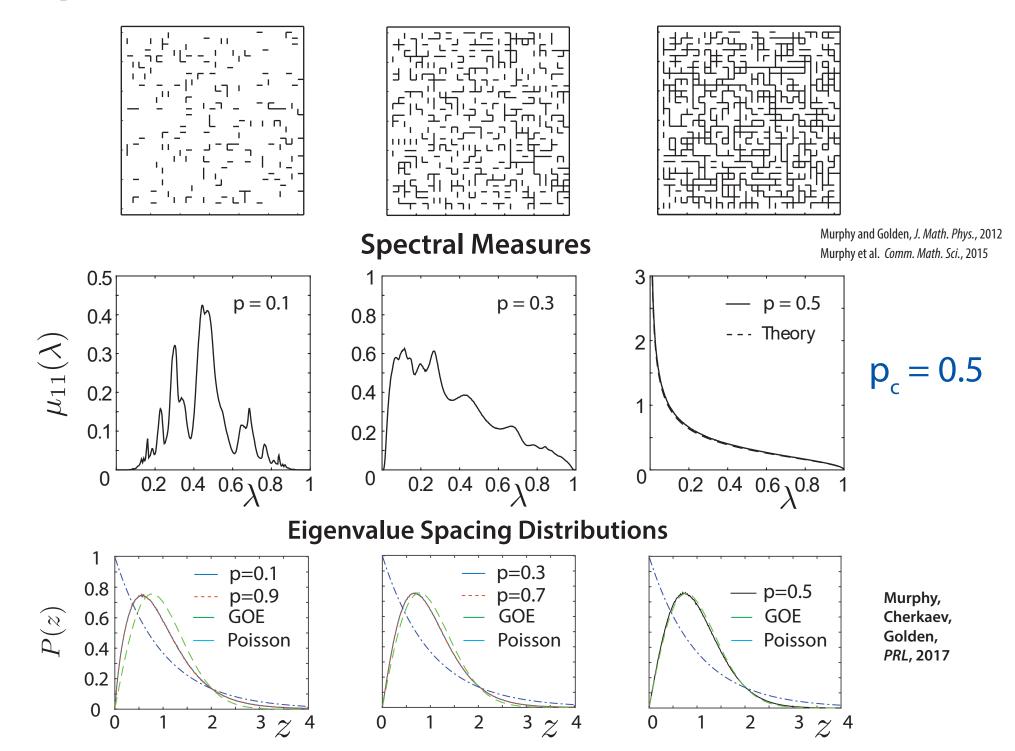
$$\mu(\lambda) = \sum_{i} \alpha_{i} \, \delta(\lambda - \lambda_{i})$$

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

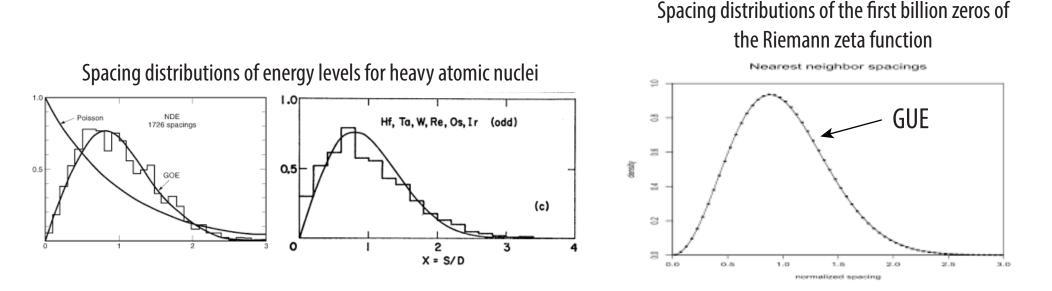


Eigenvalue Statistics of Random Matrix Theory

Wigner (1951) and Dyson (1953) first used random matrix theory (RMT) to describe quantized energy levels of heavy atomic nuclei.

 $\begin{bmatrix} \mathbf{N} \end{bmatrix}_{ij} \sim N(0,1), \qquad \mathbf{A} = (\mathbf{N} + \mathbf{N}^{\mathsf{T}})/2 \qquad \textbf{Gaussian orthogonal ensemble (GOE)}$ $\begin{bmatrix} \mathbf{N} \end{bmatrix}_{ij} \sim N(0,1) + iN(0,1), \qquad \mathbf{A} = (\mathbf{N} + \mathbf{N}^{\mathsf{T}})/2 \qquad \textbf{Gaussian unitary ensemble (GUE)}$

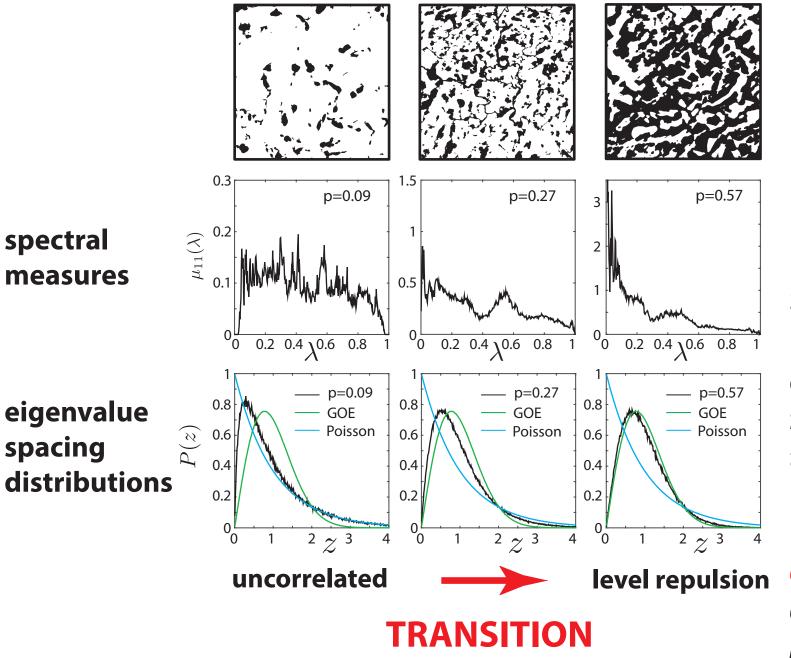
Short range and long range correlations of eigenvalues are measured by various eigenvalue statistics



RMT used to characterize **disorder-driven transitions** in mesoscopic conductors, neural networks, random graph theory, etc.

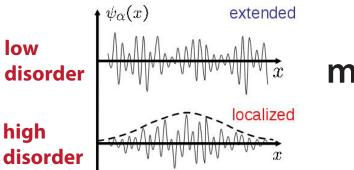
Phase transitions ~ transitions in universal eigenvalue statistics.

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 ! --

eigenvector localization and mobility edges

Inverse Participation Ratio:
$$I(\vec{v}_n) = \sum_{i=1}^N |(\vec{v}_n)_i|^4$$

Completely Localized: $I(\vec{e}_n) = 1$

Completely Extended: $I\left(\frac{1}{\sqrt{N}}\vec{1}\right) = \frac{1}{N}$

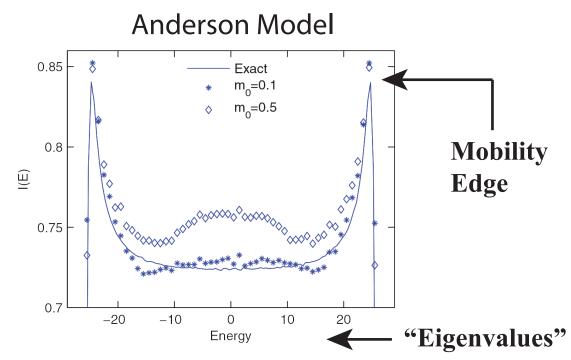
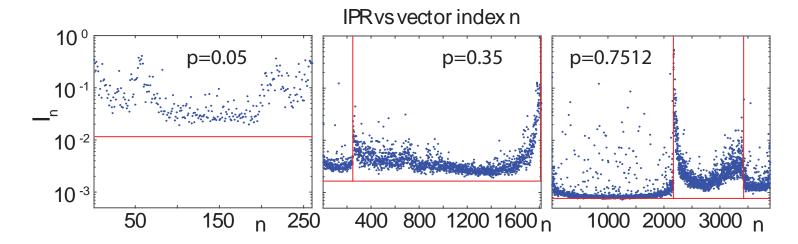
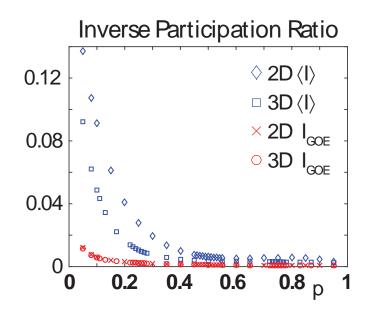


FIG. 4. (Color online) IPR for Anderson model in two dimensions with x = 6.25 (w = 50) from exact diagonalization (solid line) and from LDRG with different values of the cutoff m_0 . LDRG data are averaged over 100 runs of systems with 100 × 100 sites.

PHYSICAL REVIEW B 90, 060205(R) (2014)

Localization properties of eigenvectors in random resistor networks





$$I_n = \sum_i (\vec{v}_n)_i^4$$

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

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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



advection enhanced diffusion

effective diffusivity

sea ice floes diffusing in ocean currents 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 $\,\vec{u}\,$

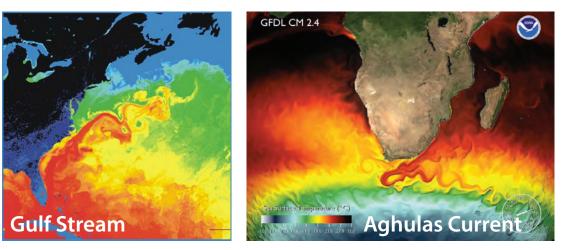
$$\frac{\partial T}{\partial t} + \vec{u} \cdot \vec{\nabla} T = \kappa_0 \Delta T$$
$$\vec{\nabla} \cdot \vec{u} = 0$$
$$homogenize$$
$$\frac{\partial \overline{T}}{\partial t} = \kappa^* \Delta \overline{T}$$

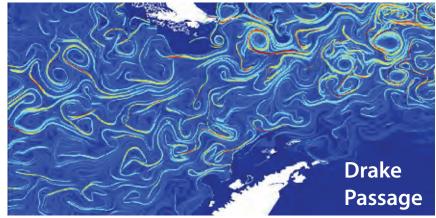
κ^{*} effective diffusivity

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, 2018

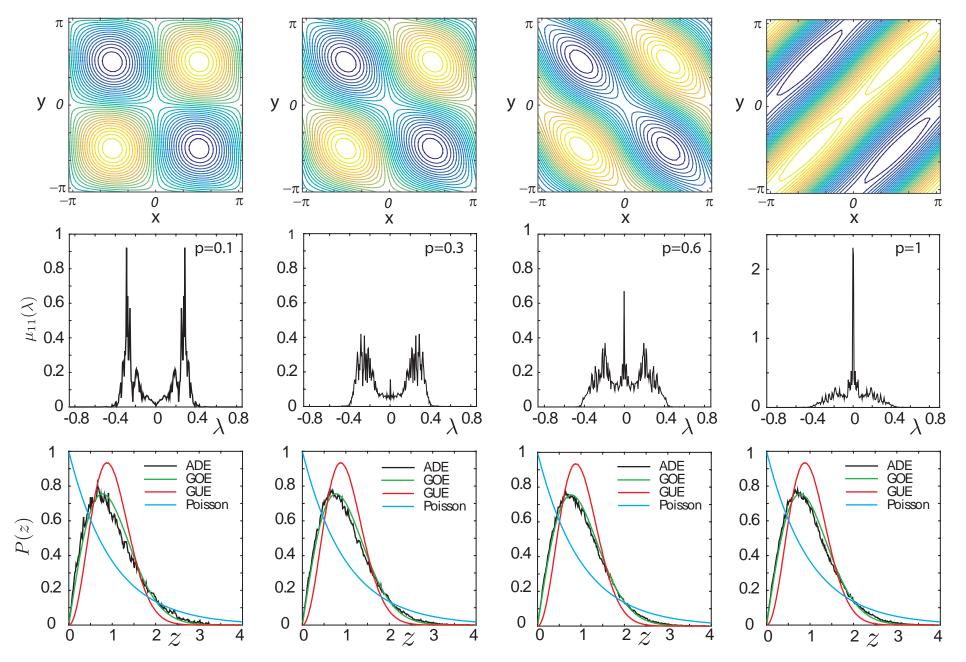






Spectral measures and eigenvalue spacings for cat's eye flow

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



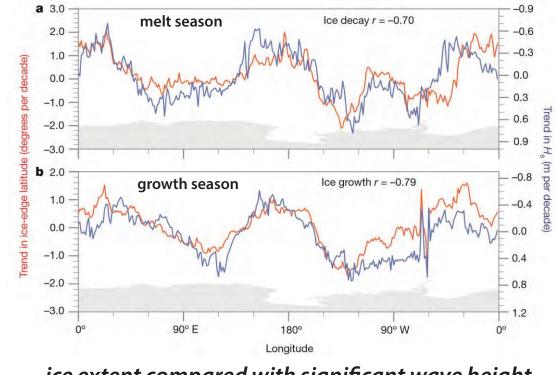
Murphy, Cherkaev, Zhu, Xin, Golden

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

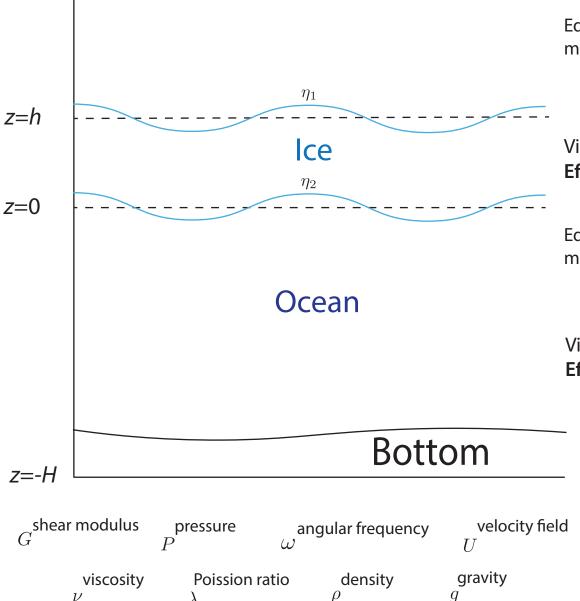




ice extent compared with significant wave height

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 $\nu_e = \nu + iG/\rho\omega$

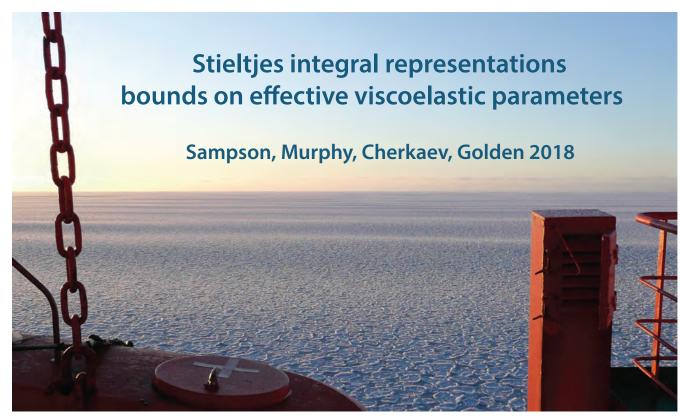
Equations of $\frac{\partial U}{\partial t} = -\frac{1}{\rho}\nabla P + \nu_e \nabla^2 U + g$

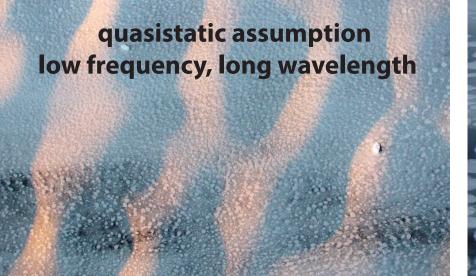
Viscoelastic thin beam (Mosig et al. 2015) Effective Complex Shear Modulus $G_v = G - i\omega\rho\nu$

Stieltjes integral representation for effective complex viscoelastic parameter; bounds

Sampson, Murphy, Cherkaev, Golden 2018

wave propagation in the marginal ice zone







Stieltjes Integral Representation for Complex Viscoelasticity

homogenized

$$\begin{cases}
\langle \sigma_{ij} \rangle = C_{ijkl}^* \langle \epsilon_{kl} \rangle \\
\text{local} \quad \nabla \cdot \sigma = 0 \qquad \sigma_{ij} = C_{ijkl} \epsilon_{kl} & \text{Strain Field} \\
C_{ijkl} = (\nu_1 \chi + (1 - \chi)\nu_2)\lambda_s \qquad \epsilon = \frac{1}{2} [\nabla u + (\nabla u)^T] = \nabla^s u \\
\nabla \cdot ((\nu_1 \chi + (1 - \chi)\nu_2)\lambda_s; \epsilon) = 0 \qquad \epsilon = \epsilon_0 + \epsilon_f \text{ where } \epsilon_f = \nabla^s \phi \\
s = \frac{1}{1 - \frac{\nu_1}{\nu_2}} & \text{Elasticity Tensor} \\
c_{ijkl}^* = \nu^* \left(\delta_{ik} \delta_{jl} + \delta_{il} \delta_{jk} - \frac{2}{3} \delta_{ij} \delta_{kl} \right) = \nu^* \lambda_s
\end{cases}$$
RESOLVENT
$$\epsilon = \left(1 - \frac{1}{s} \Gamma \chi \right)^{-1} \epsilon_0 \qquad \Gamma = \nabla^s (\nabla \cdot \nabla^s)^{-1} \nabla \cdot \epsilon_0 \text{ avg strain} \\
\chi^*$$

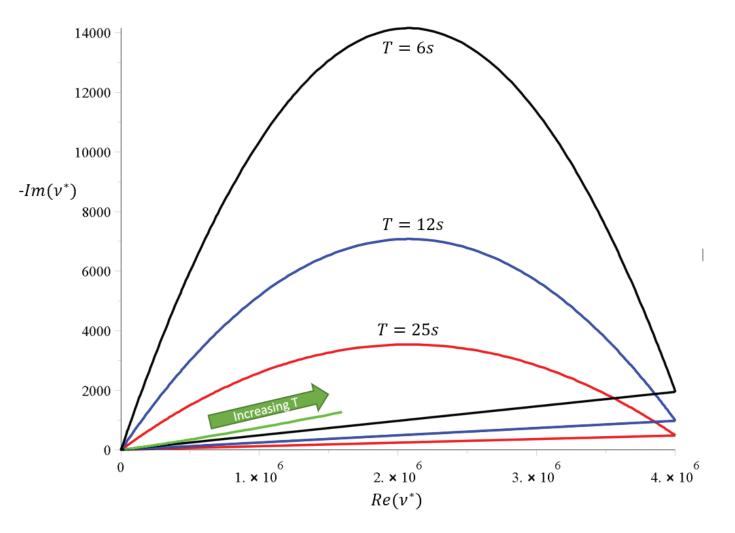
$$F(s) = 1 - \frac{v}{v_2} \qquad F(s) = \left\| \epsilon_0 \right\|^{-2} \int_{\Sigma} \frac{d\mu(n)}{s - \lambda}$$

bounds on the effective complex viscoelasticity

complex elementary bounds V_1 (fixed area fraction of floes) V_2

 $V_1 = 10^7 + i\,4875$ pancake ice

 $V_2 = 5 + i \, 0.0975$ slush / frazil



Sampson, Murphy, Cherkaev, Golden 2018

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. Statistical physics and homogenization help *link scales*, provide rigorous methods for finding effective behavior, and advance how sea ice is represented in climate models.
- 4. Random matrix theory and an unexpected Anderson transition arises in our studies of percolation in sea ice structures.
- 5. Our research will help to improve projections of climate change and the fate of the Earth sea ice packs.

THANK YOU

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Division of Mathematical Sciences Division of Polar Programs

Office of Naval Research

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