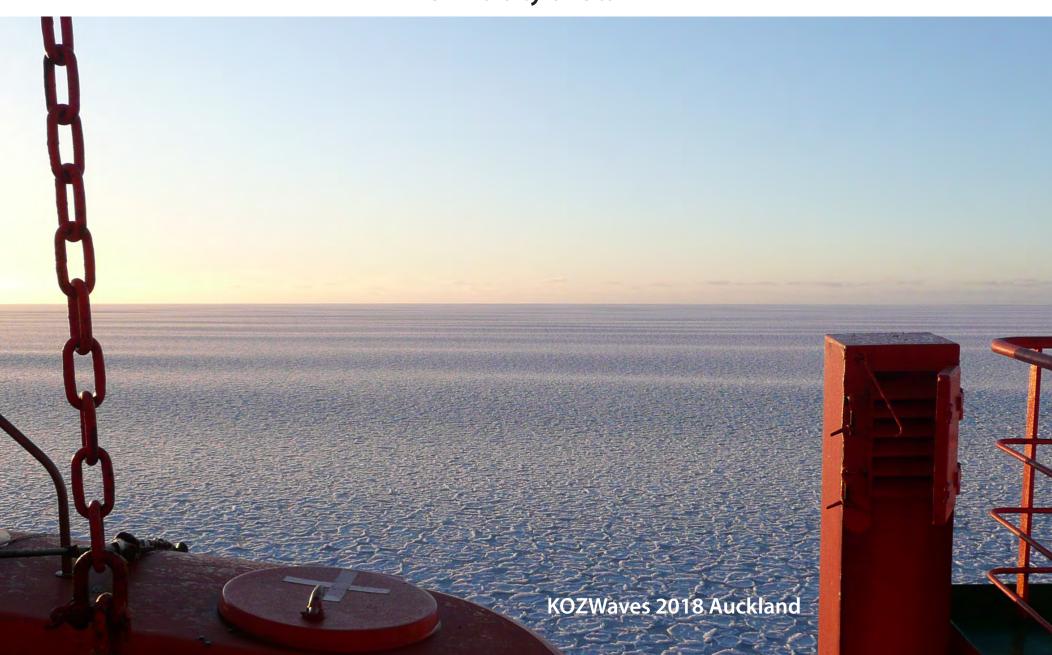
#### Homogenization, Anderson Transitions, and Waves in Sea Ice

Kenneth M. Golden
Department of Mathematics
University of Utah



## sea ice is a multiscale composite

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



millimeters



pancakes

eters centimeters





meters



kilometers

ice floes

#### 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!

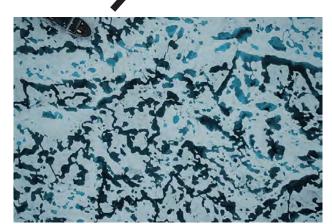
- 1. Sea ice microphysics and fluid transport homogenization and percolation theory
- 2. EM monitoring of sea ice, analytic continuation method random matrix theory and Anderson transitions
- 3. Extension of ACM to advection diffusion, waves in sea ice
  Stieltjes integral representations, spectral measures

# How do scales interact in the sea ice system?



basin scale grid scale albedo

km scale melt ponds



Linking



**Linking Scales** 



km scale melt ponds

**Scales** 



meter scale snow topography

mm
scale
brine
inclusions

## sea ice microphysics

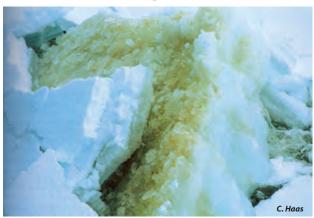
fluid transport

## 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

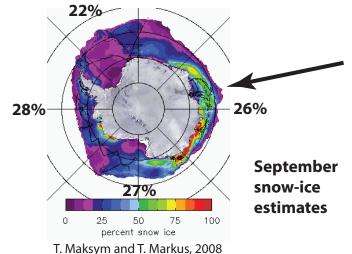


nutrient flux for algal communities





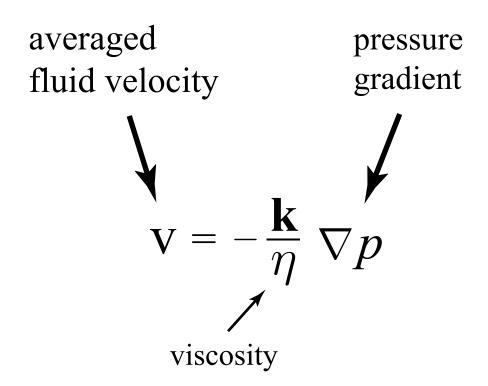




Antarctic surface flooding and snow-ice formation

- evolution of salinity profiles
- ocean-ice-air exchanges of heat, CO<sub>2</sub>

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

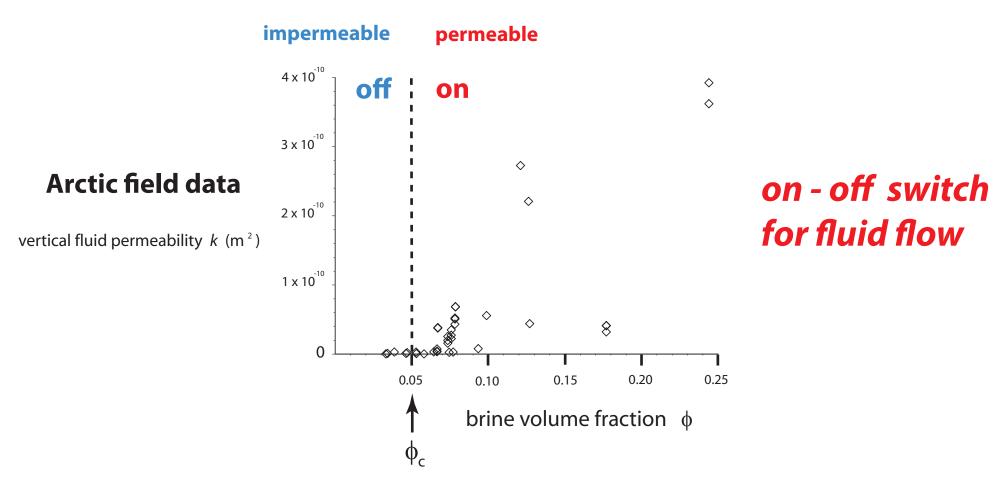


**k** = fluid permeability tensor example of *homogenization* 

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



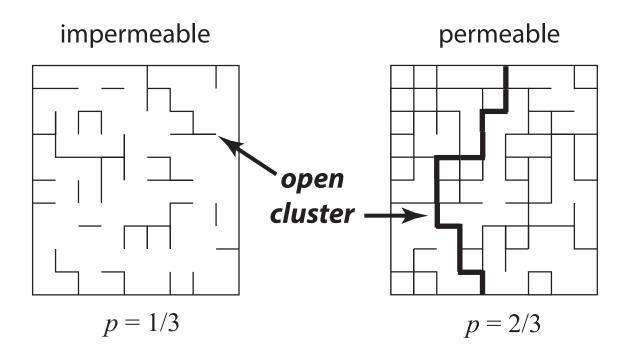
critical brine volume fraction  $\phi_c \approx 5\%$   $\longrightarrow$   $T_c \approx -5^{\circ} \text{C}$ ,  $S \approx 5 \text{ ppt}$ 

### RULE OF FIVES

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

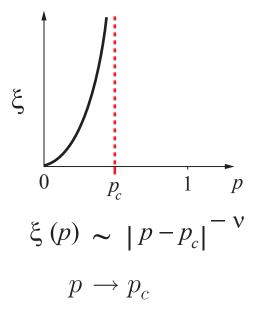
$$p_c = 1/2$$
 for  $d = 2$ 

smallest p for which there is an infinite open cluster

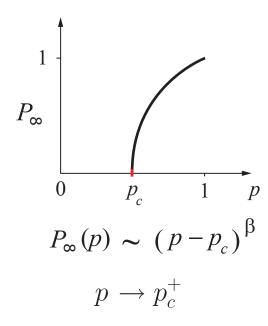
### order parameters in percolation theory

#### geometry

## correlation length characteristic scale of connectedness

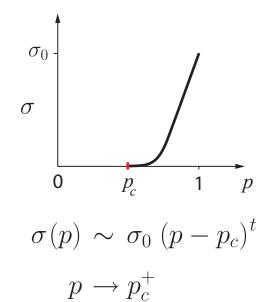


infinite cluster density probability the origin belongs to infinte cluster



#### transport

effective conductivity or fluid permeability



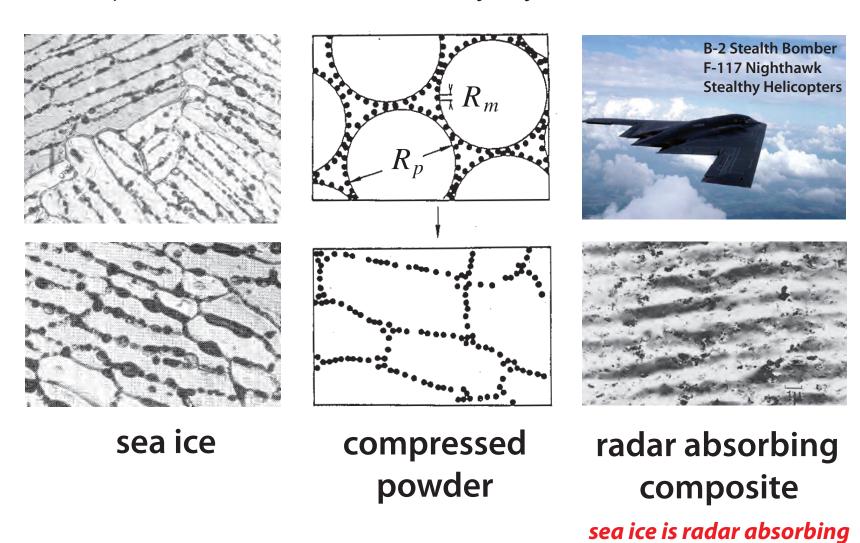
#### UNIVERSAL critical exponents for lattices -- depend only on dimension

 $1 \le t \le 2$  (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





rigorous bounds percolation theory hierarchical model network model

field data

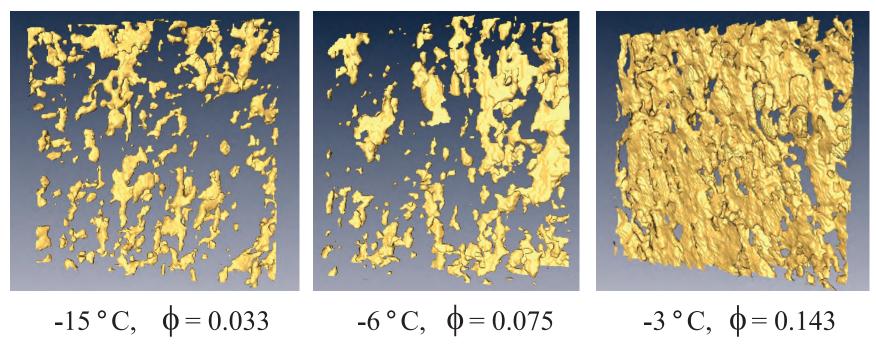
X-ray tomography for brine inclusions

unprecedented look at thermal evolution of brine phase and its connectivity

micro-scale controls macro-scale processes

### brine connectivity (over cm scale)

 $8 \times 8 \times 2 \text{ mm}$ 



#### X-ray tomography confirms percolation threshold

3-D images pores and throats



3-D graph 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

## Remote sensing of sea ice











sea ice thickness ice concentration

#### **INVERSE PROBLEM**

Recover sea ice properties from electromagnetic (EM) data

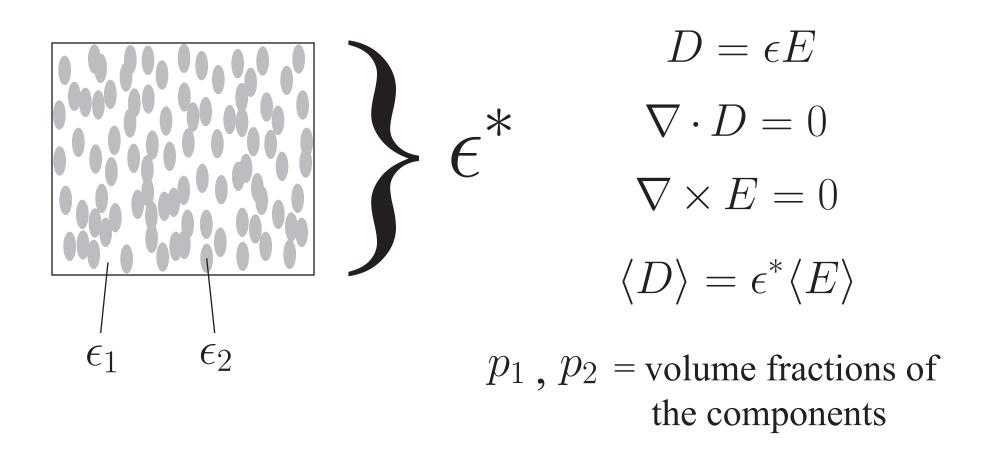
٤\*

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



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

**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.

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

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

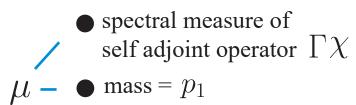


recover brine volume fraction, connectivity, etc.

## Stieltjes integral representation

#### separates geometry from parameters

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



 higher moments depend on *n*-point correlations

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

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

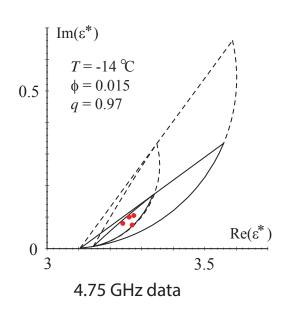
$$E = (s + \Gamma \chi)^{-1} e_k$$

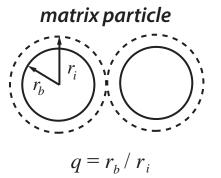
## $\Gamma \chi$ : microscale $\rightarrow$ macroscale

 $\Gamma \chi$  links scales

#### forward and inverse bounds on the complex permittivity of sea ice

#### forward bounds





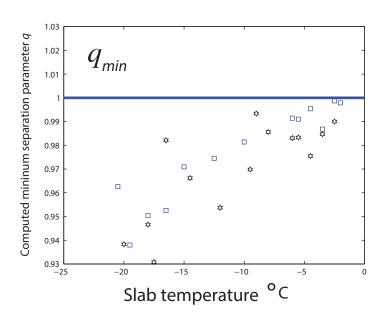
0 < q < 1

Golden 1995, 1997 Bruno 1991

## inverse bounds and recovery of brine porosity

Gully, Backstrom, Eicken, Golden Physica B, 2007

#### inverse bounds



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

## 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**

#### **HUMAN BONE**

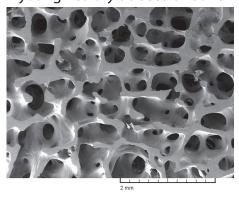


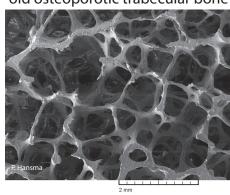


spectral characterization of porous microstructures in human bone

young healthy trabecular bone

old osteoporotic trabecular bone





reconstruct spectral measures from complex permittivity data

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!

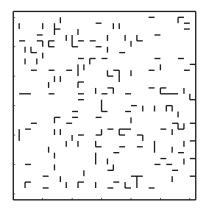
#### direct calculation of spectral measure

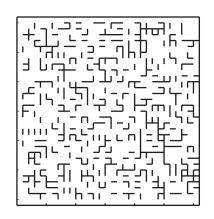
- 1. Discretization of composite microstructure gives lattice of 1's and 0's (random resistor network).
- 2. The fundamental operator  $\chi \Gamma \chi$  becomes a random matrix depending only on the composite geometry.
- 3. Compute the eigenvalues  $\lambda_i$  and eigenvectors of  $\chi \Gamma \chi$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

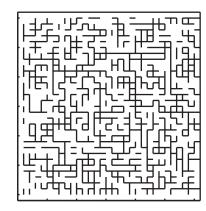
$$\mu(\lambda) = \sum_{i} \alpha_{i} \, \delta(\lambda - \lambda_{i})$$

Dirac point measure (Dirac delta)

#### **Spectral statistics for 2D random resistor network**



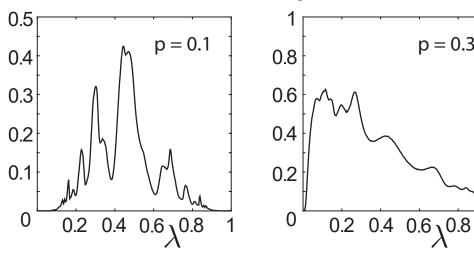


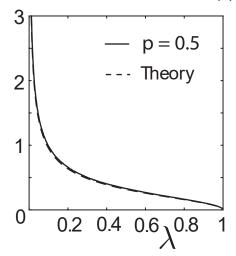


Murphy and Golden, J. Math. Phys., 2012 Murphy et al. Comm. Math. Sci., 2015



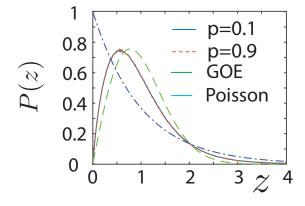
p = 0.3



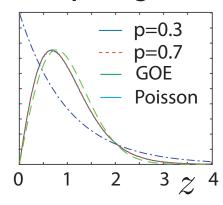


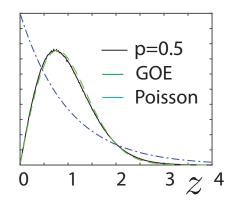
 $p_{c} = 0.5$ 

#### **Eigenvalue Spacing Distributions**



 $\mu_{11}(\lambda)$ 





Murphy, Cherkaev, Golden, PRL, 2017

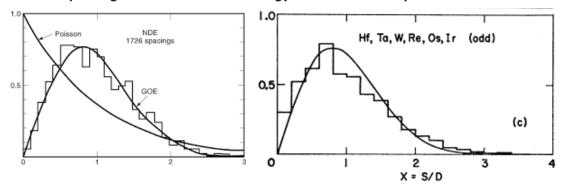
#### **Eigenvalue Statistics of Random Matrix Theory**

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

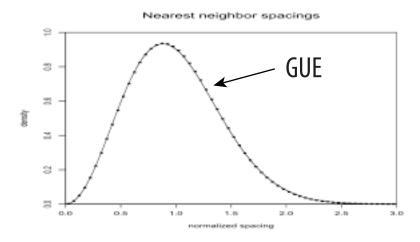
$$[N]_{ij} \sim N(0,1),$$
  $A = (N+N^T)/2$  Gaussian orthogonal ensemble (GOE)  $[N]_{ij} \sim N(0,1) + iN(0,1),$   $A = (N+N^T)/2$  Gaussian unitary ensemble (GUE)

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

Spacing distributions of energy levels for heavy atomic nuclei



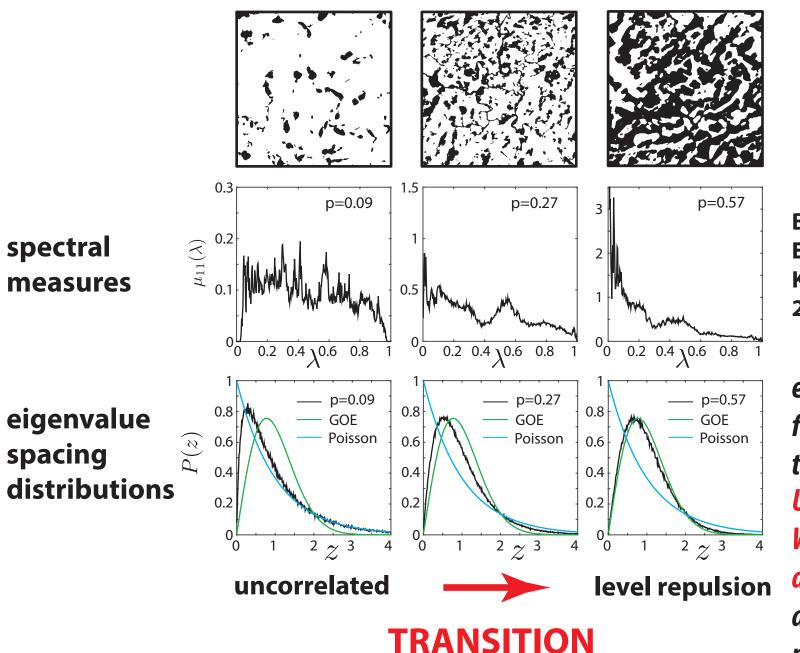
Spacing distributions of the first billion zeros of the Riemann zeta function



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

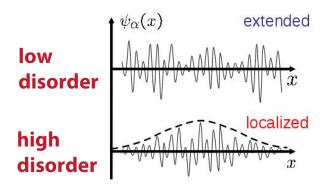
Phase transitions ~ transitions in universal eigenvalue statistics.

#### **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

PERCOLATION TRANSITION



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

#### **Anderson Model**

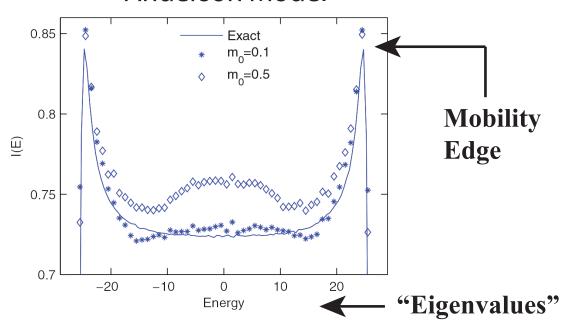
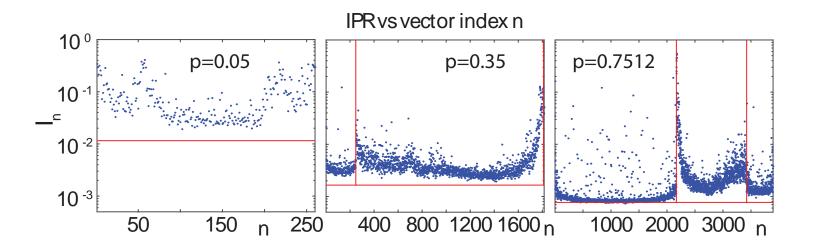
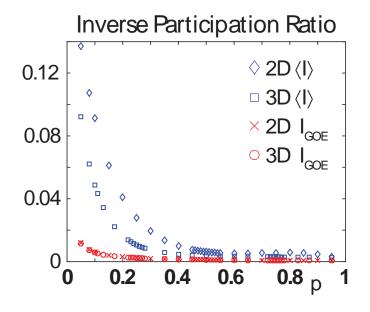


FIG. 4. (Color online) IPR for Anderson model in two dimensions with x = 6.25 (w = 50) from exact diagonalization (solid line) and from LDRG with different values of the cutoff  $m_0$ . LDRG data are averaged over 100 runs of systems with  $100 \times 100$  sites.

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

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

#### **PROCEEDINGS A**

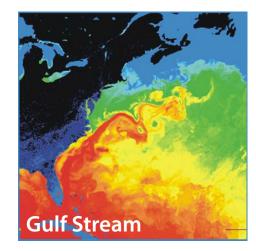


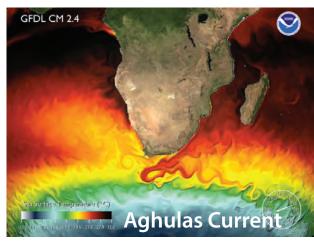
An invited review commemorating 350 years of scientific publishing at the Royal Society A method to distinguish between different types of sea ice using remote sensing techniques A computer model to determine how a human should walk so as to expend the least energy



## 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





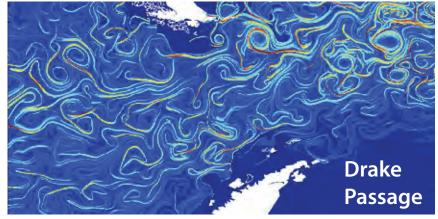
advection diffusion equation with a velocity field  $ec{u}$ 

 $\kappa^*$  effective diffusivity

#### Stieltjes integral for $\kappa^*$ with spectral measure

Avellaneda and Majda, PRL 89, CMP 91

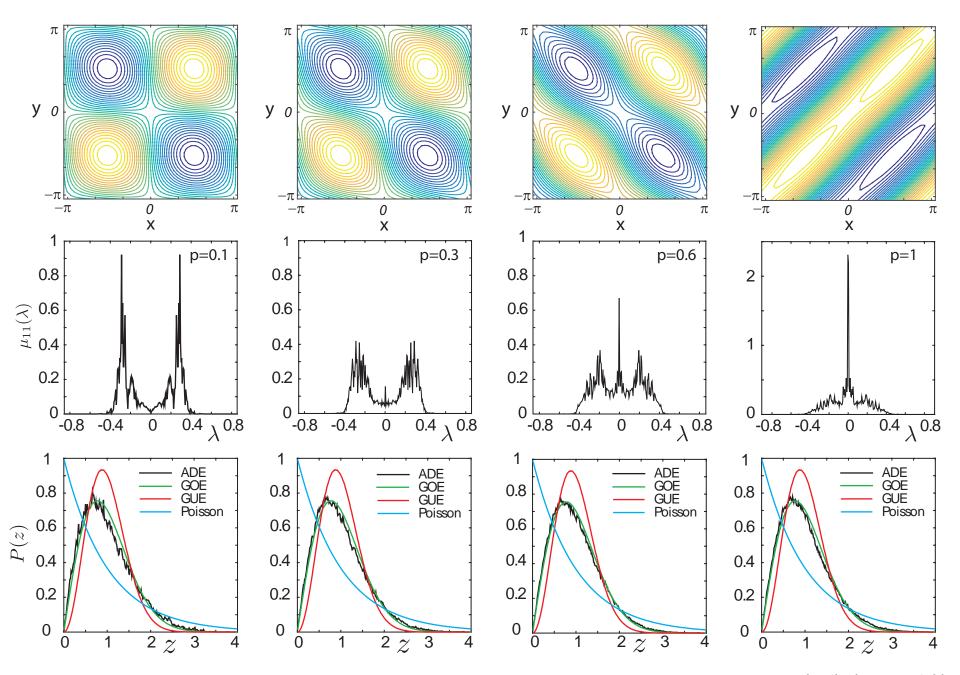
Murphy, Cherkaev, Xin, Zhu, Golden, *Ann. Math. Sci. Appl.* 2017 Murphy, Cherkaev, Zhu, Xin, Golden, 2018





#### Spectral measures and eigenvalue spacings for cat's eye flow

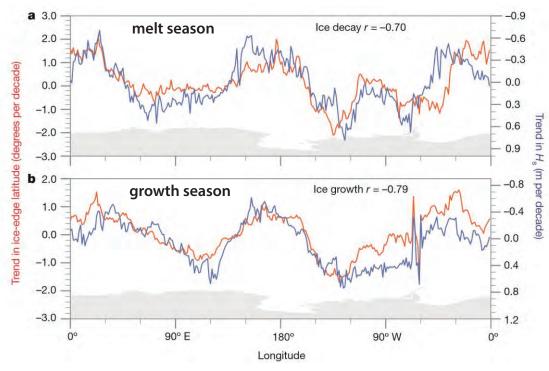
 $H(x,y) = \sin(x)\sin(y) + A\cos(x)\cos(y), \quad A \sim U(-p,p)$ 



## 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.
- large waves break sea ice much farther from the ice edge than would be predicted by the commonly assumed exponential decay

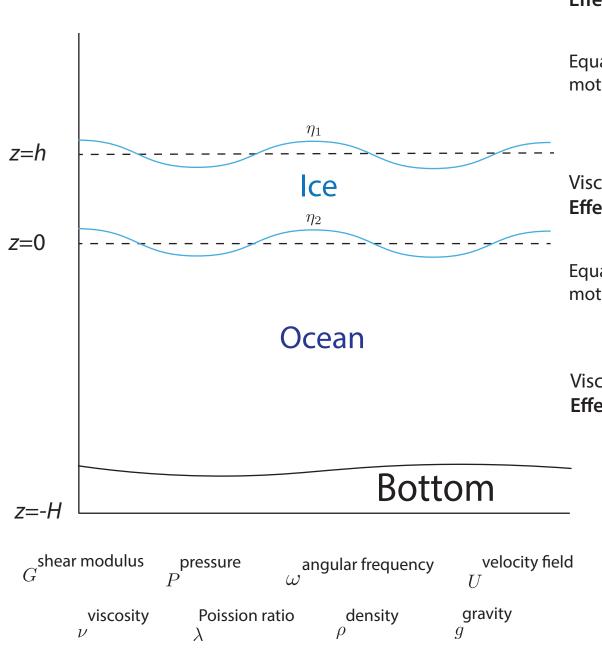




ice extent compared with significant wave height

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

### Two Layer Models and Effective Parameters



Viscous fluid layer (Keller 1998) **Effective Viscosity**  $\nu$ 

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

Viscoelastic fluid layer (Wang-Shen 2010)

Effective Complex Viscosity  $v_e = \nu + iG/\rho\omega$ 

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

Viscoelastic thin beam (Mosig et al. 2015)

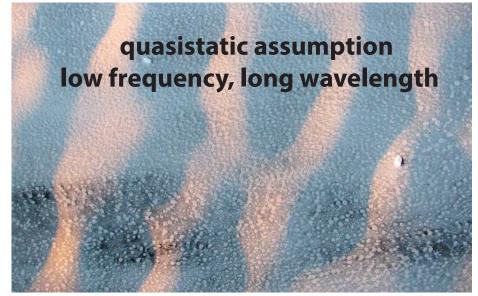
Effective Complex Shear Modulus  $G_v = G - i\omega\rho\nu$ 

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** $\langle \sigma_{ij} \rangle = C^*_{ijkl} \langle \epsilon_{kl} \rangle$

$$\langle \sigma_{ij} \rangle = C_{ijkl}^* \langle \epsilon_{kl} \rangle$$

$$\nabla \cdot \sigma = 0$$

**local** 
$$\nabla \cdot \sigma = 0$$
  $\sigma_{ij} = C_{ijkl} \epsilon_{kl}$ 

$$C_{ijkl} = (\nu_1 \chi + (1 - \chi)\nu_2)\lambda_s$$

$$\epsilon = \frac{1}{2} [\nabla u + (\nabla u)^T] = \nabla^s u$$

$$\nabla \cdot \left( (\nu_1 \chi + (1 - \chi) \nu_2) \lambda_s : \epsilon \right) = 0$$

$$\epsilon = \epsilon_0 + \epsilon_f$$
 where  $\epsilon_f = \nabla^s \phi$ 

$$s = \frac{1}{1 - \frac{\nu_1}{\nu_2}}$$

**Elasticity Tensor** 

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

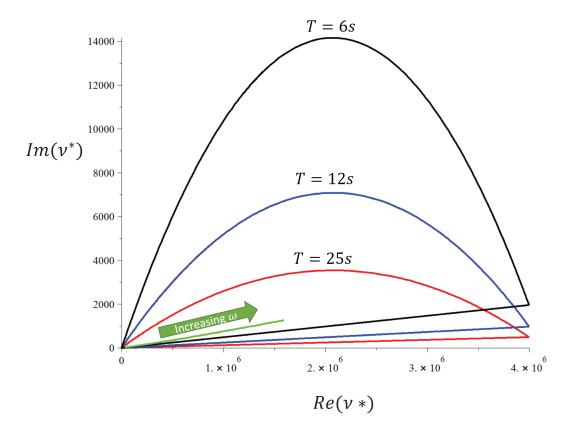
**RESOLVENT** 
$$\epsilon = \left(1 - \frac{1}{s}\Gamma\chi\right)^{-1}\epsilon_0$$
  $\Gamma = \nabla^s(\nabla\cdot\nabla^s)^{-1}\nabla\cdot$   $\epsilon_0$  avg strain

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

$$F(s) = 1 - \frac{\nu^*}{\nu_2}$$

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

### bounds on the effective complex viscoelasticity



## complex elementary bounds (fixed area fraction of floes)

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

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

T = 25s

2. × 10<sup>6</sup>

Re(v\*)

3. × 10<sup>6</sup>

4. × 10<sup>6</sup>

 $Im(v^*)$ 1000 - Increasing 0

1. × 10<sup>6</sup>

3000

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**

#### **National Science Foundation**

Division of Mathematical Sciences

Division of Polar Programs



Arctic and Global Prediction Program

Applied and Computational Analysis Program

















