Herglotz Functions in the Mathematics of Sea Ice and Other Composites

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Alison Kohout September 2012

Institute Mittag Leffler Workshop Herglotz-Nevanlinna functions and their applications

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ANTARCTICA

southern cryosphere

Weddell Sea

East Antarctic Ice Sheet

West Antarctic Ice Sheet

Ross Sea

sea ice

SEA ICE covers 7 - 10% 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

Stroeve et al., GRL, 2007



YEAR

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



What is this talk about?

Using the mathematics of composite materials and statistical physics to study sea ice structures and processes ... to improve projections of climate change.

A tour of Herglotz functions and how they arise in the study of composites, and sea ice in particular

1. Fluid flow through sea ice, percolation

- 2. Homogenization for two phase composites remote sensing, inversion, spectral measures
- 3. Stieltjes representations for advection diffusion, polycrystals, ocean waves in the marginal ice zone
- 4. Herglotz functions and the Ising model
- 5. Multiphase media and the polydisc
- 6. Arctic and Antarctic field experiments



Global Climate Models

Climate models are systems of partial differential equations (PDE) derived from the basic laws of physics, chemistry, and fluid motion.

They describe the state of the ocean, ice, atmosphere, land, and their interactions.

The equations are solved on 3-dimensional grids of the air-ice-ocean-land system (with horizontal grid size ~ 50 km), using very powerful computers.

key challenge :

incorporating sub - grid scale processes

linkage of scales



sea ice components of GCM's

What are the key ingredients -- or *governing equations* that need to be solved on grids using powerful computers?

1. Ice thickness distribution g(x, y, h, t) evolution equation dynamics

$$\frac{Dg}{Dt} = -g\nabla \cdot \mathbf{u} + \Psi(g) - \frac{\partial}{\partial h}(fg) + \mathcal{L}$$

nonlinear PDE incorporating ice velocity field ice growth and melting mechanical redistribution - ridging and opening



2. Conservation of momentum, stress vs. strain relation (Hibler 1979)

(Maykut and Untersteiner 1971)

$$mrac{D{f u}}{Dt}=-mf{f k} imes{f u}+{m au}_a+{m au}_o-mg
abla H+{f F}$$
 F=ma for sea ice dynamics

3. Heat equation of sea ice and snow

thermodynamics

+ balance of radiative and thermal fluxes on interfaces

$$\frac{\partial T}{\partial t} + \mathbf{u} \cdot \nabla T = \nabla \cdot k(T) \,\nabla T$$

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

HOMOGENIZE as $\epsilon \to 0$

Stokes equations for fluid velocity \mathbf{v}^{ϵ} , pressure p^{ϵ} , force **f**:



$$\nabla p^{\epsilon} - \epsilon^2 \eta \Delta \mathbf{v}^{\epsilon} = \mathbf{f}, \quad x \in \mathcal{P}_{\epsilon}$$
$$\nabla \cdot \mathbf{v}^{\epsilon} = 0, \quad x \in \mathcal{P}_{\epsilon}$$
$$\mathbf{v}^{\epsilon} = 0, \quad x \in \partial \mathcal{P}_{\epsilon}$$
$$\eta = \text{fluid viscosity}$$

via two-scale expansion

MACROSCOPIC EQUATIONS $\mathbf{v}^{\epsilon} \rightarrow \mathbf{v}$, $p^{\epsilon} \rightarrow p$ as $\epsilon \rightarrow 0$ Darcy's law $\mathbf{v} = -\frac{1}{\eta} \mathbf{k} \nabla p$, $x \in \Omega$ $\mathbf{k}(x) =$ effective fluid
permeability
tensor $(\mathbf{f} = \mathbf{0})$ $\nabla \cdot \mathbf{v} = 0$, $x \in \Omega$ $\mathbf{k}(x) =$ effective fluid
permeability
tensor

[Keller '80, Tartar '80, Sanchez-Palencia '80, J. L. Lions '81, Allaire '89, '91,'97]

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



 $\mathbf{k} =$ fluid permeability tensor

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

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

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

HOMOGENIZATION



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

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

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

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



 p_1 , p_2 = volume fractions of brine and ice

Effective complex permittivity of a two phase composite in the quasistatic (long wavelength) limit



the components

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

ocean swells propagating through a vast field of pancake ice

HOMOGENIZATION: long wave sees an effective medium, not individual floes



Analytic continuation method for bounding complex ϵ^*

Bergman (1978), Milton (1979), Golden and Papanicolaou (1983)

$$m(h) = \frac{\epsilon^*}{\epsilon_2} \left(\frac{\epsilon_1}{\epsilon_2}\right) \qquad h = \frac{\epsilon_1}{\epsilon_2}$$



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.

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

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

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

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

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

Golden and Papanicolaou, Comm. Math. Phys. 1983
using the Stieltjes integral representation to obtain bounds "linear programming" Golden and Papanicolaou, CMP 1983 M_1 = the set of positive Borel measures on [0,1], compact, convex $F_s(\mu): M_1 \longrightarrow \mathbb{C}$ linear functional extremal values (bounds) are images of extreme points of M_1 $\frac{\mu_0}{s-z^*}$ **Dirac point measures**

higher order bounds -- iterated fractional linear transformations

Rakar 1060

$$F_1(s) = \frac{1}{\mu_0} - \frac{1}{sF(s)} \qquad \longleftarrow \qquad \begin{array}{c} \text{Milton 1981} \\ \text{Bergman 1982} \\ \text{Felderhof 1984} \\ \text{Golden 1986} \end{array}$$

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*

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

HUMAN BONE





SEA ICE

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

Golden, Murphy, Cherkaev, J. Biomechanics 2011

spectral characterization of porous microstructures in bone

Golden, Murphy, Cherkaev, J. Biomechanics 2011

(a) young healthy trabecular bone



nm

(c) spectral measure - young



(b) old osteoporotic trabecular bone



(d) spectral measure - old

using regularized inversion scheme

reconstruction of spectral measures from complex

permittivity data

EM monitoring of osteoporosis

loss of bone connectivity

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

reconstruction of spectral measures from simulated complex permittivity data



regularized inversion scheme

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

Continuum composite



Spectral measures of

 $\chi_1 \Gamma \chi_1$

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



Integro-differential projection operator $\Gamma = \vec{\nabla} (\Delta^{-1}) \vec{\nabla} \cdot$

Point-wise indicator function

 χ_1

Resolvent representation of electric field

$$\chi_1 \vec{E} = sE_0(sI - \chi_1 \Gamma \chi_1)^{-1} \chi_1 \vec{e}_k$$

Integral representation

$$F(s) = \int_0^1 \frac{d\mu(\lambda)}{s - \lambda}$$

Projection matrix $\Gamma = \nabla (\nabla^{\mathrm{T}} \nabla)^{-1} \nabla^{\mathrm{T}}$

Diagonal projection matrix

 χ_1

Series representation of electric field

$$\chi_1 \vec{E} = sE_0 \sum_j \frac{\vec{v}_j \cdot \chi_1 \vec{e}_k}{s - \lambda_j} \, \vec{v}_j$$

Series representation

$$F(s) = \sum_{j} \frac{(\vec{v}_j \cdot \chi_1 \vec{e}_k)^2}{s - \lambda_j}$$

Spectral Measures for Random Resistor Networks

2-D



spectral gaps collapse at the percolation transitions

Murphy and Golden, J. Math. Phys. (2012)

Spectral Measures for Sea Ice Structures: Brine Inclusions



N. B. Murphy, C. Hohenegger, C. S. Sampson, D. K. Perovich, H. Eicken, E. Cherkaev, B. Alali, and K. M. Golden

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

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.

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

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.

Transition in Eigenvalue Correlations

	5			
$P(z) = \exp(-z)$	P(z	$) \approx \frac{\pi z}{2} \exp(-\pi z^2)$	/4) Wigner sur	
genvalue Spacing Distributio	on Eiger	Eigenvalue Spacing Distribution		
Poisson		GOE	Picket	
Spectra		Spectra	Fence	
	Connectedness			
/	nase transition			
	\longrightarrow			
	LEVEL			
	REPULSION			
Uncorrelated		Highly	Completely	
		Correlated	Correlated	

Spectral computations for Arctic sea ice pack



Spectral statistics for 2D random resistor network





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





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

Homogenization for composite materials



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



Mathematical formulation for composite materials



$$\vec{\nabla} \cdot \vec{J} = 0, \quad \vec{\nabla} \times \vec{E} = 0, \quad \vec{J} = \sigma \vec{E}, \qquad \vec{E} = \vec{\nabla} \phi + \vec{e}_k, \quad \langle \vec{E} \rangle = \vec{e}_k$$

Polycrystalline material

Local conductivity

$$\sigma = R \operatorname{diag}(\sigma_1, \sigma_2, \sigma_2) R^T$$
$$= \sigma_1 X_1 + \sigma_2 X_2$$

 $X_2 = I - X_1$



Continuum composite



Discrete composite



Random Rotation Matrix

Integral representations for bulk transport coefficients

The *effective conductivity* is defined in terms of the *system energy*

 $\langle \vec{J} \cdot \vec{E} \rangle = \sigma^* E_0^2, \quad \sigma^* = \sigma_{kk}^*, \quad \langle \vec{E} \rangle = E_0 \vec{e}_k$

This defines a *homogeneous medium* which behaves *macroscopically* and *energetically* just like the given inhomogeneous medium.

$$\begin{split} \vec{\nabla} \times \vec{E} &= 0, \quad \vec{\nabla} \cdot \vec{J} = 0, \quad \longrightarrow \quad X_1 \vec{E} = s E_0 (s I - X_1 \Gamma X_1)^{-1} X_1 \vec{e}_k \\ \vec{J} &= \sigma \vec{E}, \quad \sigma = \sigma_1 X_1 + \sigma_2 X_2 \\ \end{split}$$

$$\sigma^*/\sigma_2 = 1 - \langle X_1 \vec{E} \cdot \vec{e}_k \rangle / (sE_0) = 1 - F(s)$$

Gully et al., *Proc. Roy. Soc. A* 2015 Murphy, Cherkaev, Golden 2016

$$F(s) = \int_0^1 \frac{d\mu(\lambda)}{s - \lambda}$$

• spectral measure of the self-adjoint operator $X_1 \Gamma X_1$ • mass = average orientation • higher moments depend on *n*-point correlations 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



two scale homogenization for polycrystalline sea ice



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

Spectral measures for uniaxial polycrystalline media



• Random crystallographic orientations angles θ measured from the vertical direction, uniformly distributed $\theta \sim U(-\delta \pi/2, \delta \pi/2)$

Murphy, Cherkaev, Golden, 2017

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







Stieltjes integral for κ^* with spectral measure

composites

Golden and Papanicolaou, CMP 1983

$$\frac{\epsilon^*}{\epsilon_2} = 1 - \int_0^1 \frac{d\mu(\lambda)}{s - \lambda}$$
$$s = \frac{1}{1 - \epsilon_1 / \epsilon_2}$$

advection diffusion

Avellaneda and Majda, PRL 89, CMP 91

$$\frac{\kappa^*}{\kappa_0} = 1 - \int_0^\infty \frac{d\rho(z)}{t-z}$$

 $t = -1/\xi^2, \ \xi = P\acute{e}clet number$

• computations of spectral measures and effective diffusivity for model flows

 $\mathbf{i}\mathbf{\Gamma}\mathbf{H}\mathbf{\Gamma}$ $\vec{u} = \kappa_0\,\xi\,\vec{\nabla}\cdot\mathbf{H}$

- H antisymmetric vector potential *Murphy, Cherkaev, Zhu, Xin, Golden 2017*
- rigorous bounds and computations on convection enhanced thermal conductivity of sea ice *Liu, Hardenbrook, Kraitzman, Zhu, Murphy, Cherkaev, Golden, 2017*



$$\frac{\partial T}{\partial t} + \vec{u} \cdot \vec{\nabla} T = \kappa_0 \Delta T \qquad \qquad \vec{\nabla} \cdot \vec{u} = 0 \\ \vec{u} = \vec{\nabla} \cdot H$$

$$\vec{\nabla} \times \vec{E} = 0, \qquad \vec{\nabla} \cdot \vec{J} = 0, \qquad \text{Resolvent formula}$$

$$\vec{J} = \sigma \vec{E}, \qquad \sigma = \kappa_0 I + S, \qquad \longrightarrow \qquad \vec{\nabla} \phi = (\kappa_0 I - \imath \Gamma S \Gamma)^{-1} \Gamma H \vec{e}_k$$
Steady flow
$$\begin{array}{c} \text{Dynamic flow} \\ S = H \end{array} \qquad S = H + (-\Delta)^{-1} \frac{\partial}{\partial t} \end{array}$$
Projection onto curl-free fields:
$$\Gamma = -\vec{\nabla} (-\Delta)^{-1} \vec{\nabla} \cdot \vec{\nabla} \cdot \vec{v}$$

$$\kappa^*/\kappa_0 = 1 + \langle \vec{\nabla}\phi \cdot \vec{\nabla}\phi \rangle = 1 + G(\kappa_0)$$

Steady flow:Murphy, Cherkaev, Zhu, Xin, Golden 2017Dynamic flow:Murphy, Cherkaev, Xin, Zhu, Golden 2017

$$G(\kappa_0) = \int_{-\infty}^{\infty} \frac{\mathrm{d}\nu(\lambda)}{\kappa_0^2 + \lambda^2} \quad \nu - \text{ spectral measure of the self-adjoint operator } i\Gamma S\Gamma$$

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, 2016

Thermal Conduction Enhanced with Convection

- Temperature on top surface driven by atmosphere conditions
- Bottom surface in contact with sea water
- Temperature field *T* governed by a nonlinear convection-diffusion equation

$$\rho c \left(\frac{\partial T}{\partial t} + \mathbf{u} \cdot \nabla T \right) = \nabla \cdot (\kappa(T) \nabla T)$$

with a Darcy velocity ${\boldsymbol{u}}$

- Parameters:
 - $\rho = \text{bulk density}$
 - c = specific heat
 - $\kappa(T)$ = temperature dependent thermal conductivity

bounds on the effective thermal conductivity of sea ice with convection (BC flow model)



Hardenbrook, Kraitzman, Zhu, Murphy, Cherkaev, 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.

wave propagation in the marginal ice zone





Ising Model for a Ferromagnet



$$\mathcal{H}_{\omega} = -J \sum_{\langle i,j \rangle} s_i s_j - H \sum_i s_i$$

nearest neighbor Ising Hamiltonian

for any configuration $\omega \in \Omega = \{-1, 1\}^N$ of the spins $J \ge 0$



canonical partition function

$$Z_N(T, H) = \sum_{\omega \in \Omega} \exp(-\beta \mathcal{H}_{\omega}) = \exp(-\beta N f_N)$$
$$\beta = 1/kT$$

free energy per site

$$f_N(T,H) = \frac{-1}{\beta N} \log Z_N(T,H)$$



free energy
$$f(T, H) = \lim_{N \to \infty} f_N(T, H)$$

magnetization

homogenized parameter like effective conductivity

$$M(T,H) = \lim_{N \to \infty} \frac{1}{N} \left\langle \sum_{j} s_{j} \right\rangle = -\frac{\partial f}{\partial H}$$



 $\begin{array}{ll} \mbox{magnetic} & \chi(T,H) = \frac{\partial M}{\partial H} = -\frac{\partial^2 f}{\partial H^2} \geq 0 \\ \mbox{susceptibility} & \end{array}$

partition function N^{th} order polynomial in the "activity" $z = exp(-2\beta H)$

$$Z_N(z) = \sum_{n=0}^N a_n z^n, \quad a_n \ge 0$$

with remarkable property:

Theorem (Lee - Yang 1952):

If $J \ge 0$, then $Z_N = 0$ implies z lies on the unit circle |z| = 1, or equivalently, H lies on the imaginary axis (with real β).

Then the partition function can be written as

$$Z_N(z) = a_N \prod_{n=1}^N (z - z_n), \qquad |z_n| = 1$$

Then
$$f(T,H) = \frac{-1}{\beta} \int_{|t|=1} \log(z-t) d\nu(t) - 2d\beta J$$

Stieltjes integral representation for magnetization

and scaling relations for critical exponents

Baker PRL 1968

$$M(\tau) = \tau + \tau (1 - \tau^2) G(\tau^2) \qquad \tau = \tanh(\beta H)$$

$$G(\tau^2) = \int_0^\infty \frac{d\psi(y)}{1 + \tau^2 y}$$

Herglotz

$$M(T) = -\frac{\partial f}{\partial H} \sim (T_c - T)^{\beta} \qquad T \to T_c^{-1}$$

$$\chi = -\frac{\partial^2 f}{\partial H^2} \sim (T - T_c)^{-\gamma} \qquad T \to T_c^+$$

Along the critical isotherm $T = T_c$

$$M(H) \sim H^{1/\delta} \qquad H \to 0^+$$

 ψ supported in [0, S(T)] $S(T) \sim (T - T_c)^{-2\Delta}, T \to T_c^+$ $\beta = \Delta - \gamma$ $\delta = \Delta / (\Delta - \gamma)$

Baker's inequalities

$$\gamma_{n+1} - 2\gamma_n + \gamma_{n-1} \ge 0$$

critical exponents γ_n of higher derivatives of f

 $\gamma_0 = \gamma$

via analogous Herglotz structure for transport in composites, same critical analysis and scaling relations hold near percolation threshold

Golden, J. Math. Phys. 1995 Phys Rev. Lett. 1997

(Chuck Newman)

 $m(h) = \frac{\sigma^*}{\sigma_2} \qquad h = \frac{\sigma_1}{\sigma_2} \to 0 \qquad \sigma^*(p,h) \qquad \begin{array}{l} \text{effective conductivity} \\ \text{of two phase composite} \\ - \text{ lattice or continuum} \end{array}$

$$F(s) = 1 - m(h)$$
 $F(s) = \int_0^1 \frac{d\mu(w)}{s - w}$ $w = \frac{y}{y + 1}$

$$m(h) = 1 + (h-1)g(h) \qquad \qquad g(h) = \int_0^\infty \frac{d\phi(y)}{1+hy} \qquad \text{Herglotz}$$

$$\begin{split} \sigma^*(p,0) &\sim (p-p_c)^t & p \to p_c^+ & t = \Delta \\ \sigma^*(p_c,h) &\sim h^{1/\delta} & h \to 0^+ & \delta = \frac{1}{\Delta} \\ \chi(p) &= \frac{\partial m}{\partial h} \sim (p_c - p)^{-\gamma} & p \to p_c^- & h = 0 \\ \theta_h &\sim (p_c - p)^{\Delta} \text{ spectral gap} & p \to p_c^- & \gamma_{n+1} - \eta_{n+1} -$$

 $\Delta = \frac{\gamma}{2}$ lattices and continua obey same scaling relations as in stat mech

Baker's inequalities for transport

$$\gamma_{n+1} - 2\gamma_n + \gamma_{n-1} \ge 0, \ n \ge 1$$
Ising model

partition function

$$Z_N(z) = a_N \prod_{n=1}^N (z - z_n), \quad |z_n| = 1$$

free energy

$$f(T,H) = \frac{-1}{\beta} \int_{|t|=1} \log(z-t) d\nu(t)$$

order parameter

$$M(T)=-\frac{\partial f}{\partial H}$$

$$\frac{\partial^2 M}{\partial H^2} \le 0$$

G.H.S. inequality Griffiths, Hurst, Sherman *JMP* 1970

transport in composites

$$\mathcal{Z}_N(s) = \prod_{n=1}^N (s - s_n), \quad s_n \in [0, 1]$$

$$\Phi(p,s) = \int_0^1 \log(s-t) d\mu(t)$$

$$F(p,s) = \frac{\partial \Phi}{\partial s}$$

$$\frac{\partial^2 m}{\partial h^2} \le 0$$

Golden, *JMP* 1995; *PRL* 1997

Multiphase Media



$$m(h_1, h_2) = \epsilon^* / \epsilon_3$$
 $h_i = \epsilon_i / \epsilon_3$ $i = 1, 2$
 $F(s_1, s_2) = 1 - m(h_1, h_2)$
 $s_i = 1/(1 - h_i)$







Bounding the effective properties of multiphase composites

Milton, Theory of Composites, 2002

trajectory method (based on two phase, one variable theory)

Bergman, *Phys. Rep.* 1978, 1982, ... Milton, 1981, ...

polydisc representation formula (Herglotz function in several variables)

> Golden and Papanicolaou, J. Stat. Phys. 1985 Golden, J. Mech. Phys. Solids 1986

field equation recursion method

Milton, Comm. Math. Phys. (I and II) 1987

effective elasticity tensor

Ou, Complex Vars. Elliptic. Eqs. 2011

polydisc representation formula

$$f(\zeta_1, \zeta_2) = iF(s_1, s_2) \qquad f(\zeta_1, \zeta_2) : D^2 \to \{Ref > 0\}$$
$$D^2 = \{|\zeta_1| < 1\} \times \{|\zeta_2| < 1\} \qquad \zeta_j = \frac{s_j - i}{s_j + i}$$

 $f(\zeta_1, \zeta_2) = iImf(0, 0) + \frac{1}{2} \int_0^{2\pi} \int_0^{2\pi} (H_1H_2 + H_1 + H_2 - 1)\mu(dt_1, dt_2)$

 $H_1 = (e^{it_1} + \zeta_1)/(e^{it_1} - \zeta_1) \qquad H_2 = (e^{it_2} + \zeta_2)/(e^{it_2} - \zeta_2)$ $\mu \text{ is a positive Borel measure on the torus } T^2$

satisfying a Fourier condition (excludes point measures)

SUNSPOT THEOREM:

Koranyi and Pukanszky, *Am. Math. Soc. Trans.* 1963 Vladimirov and Drozhzhinov, *Mat. Zametki* 1974 Golden and Papanicolaou, *J. Stat. Phys.* 1985

conjectured extremals yield bounds



set of extremal measures = ??

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	McMurdo Sound
2011	Arctic	Barrow AK
2012	Arctic	Barrow AK
2012	Antarctic	SIPEX II
2013	Arctic	Barrow AK
2014	Arctic	Chukchi Sea



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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 2016

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

Conclusions

- 1. Summer Arctic sea ice is melting rapidly.
- 2. Fluid flow through sea ice mediates many processes of importance to understanding climate change and the response of polar ecosystems.
- 3. Homogenization and statistical physics help *link scales* and provide rigorous methods for finding effective behavior, and advance how sea ice is represented in climate models.
- 4. Herglotz functions and Stieltjes integrals provide powerful methods of homogenization for sea ice structures and processes.
- 5. Our research will help to improve projections of climate change and the fate of the Earth sea ice packs.

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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