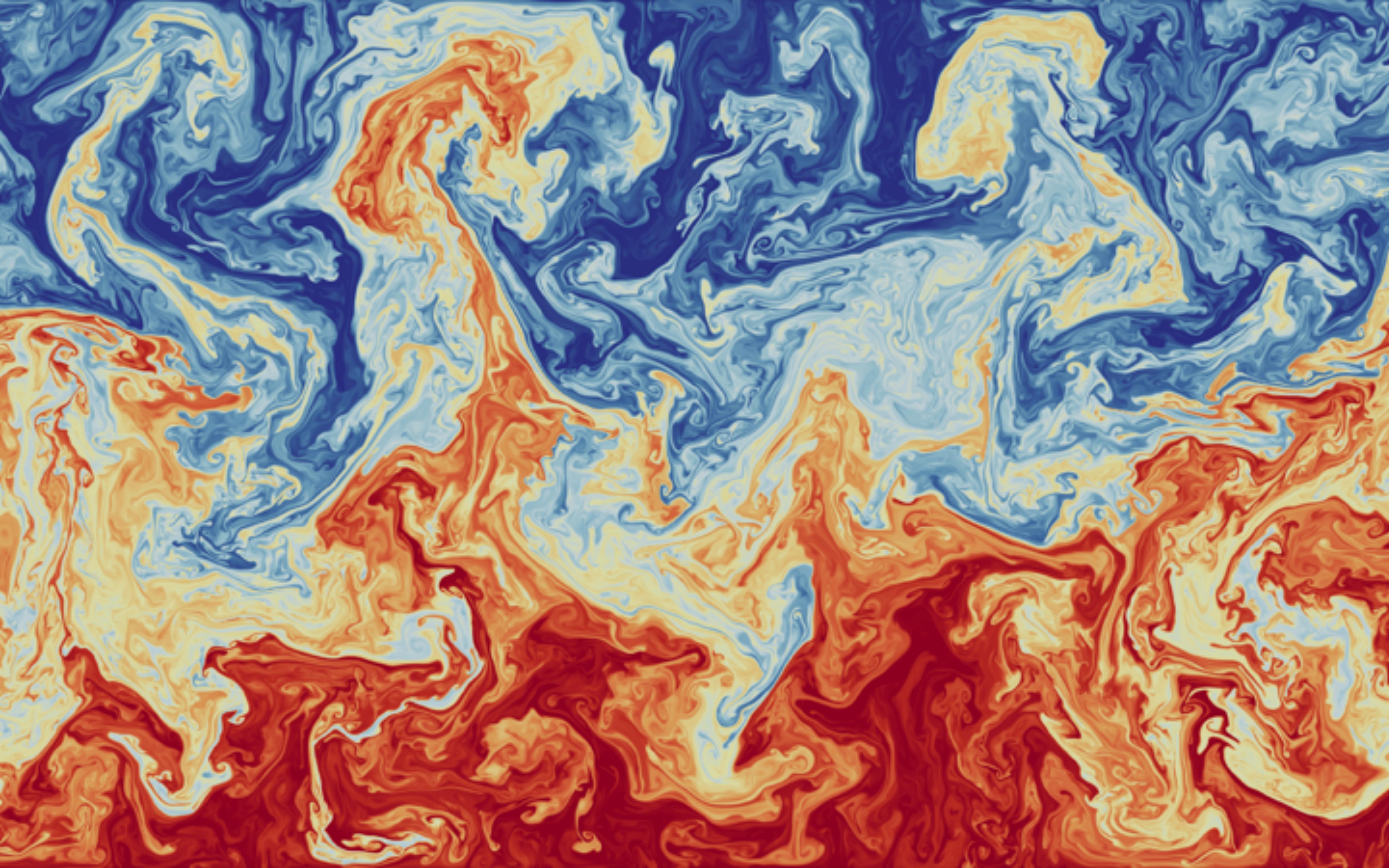


1: Introduction to spectral methods & Dedalus

Keaton Burns

CISM Udine 2023



PDEs across science & engineering



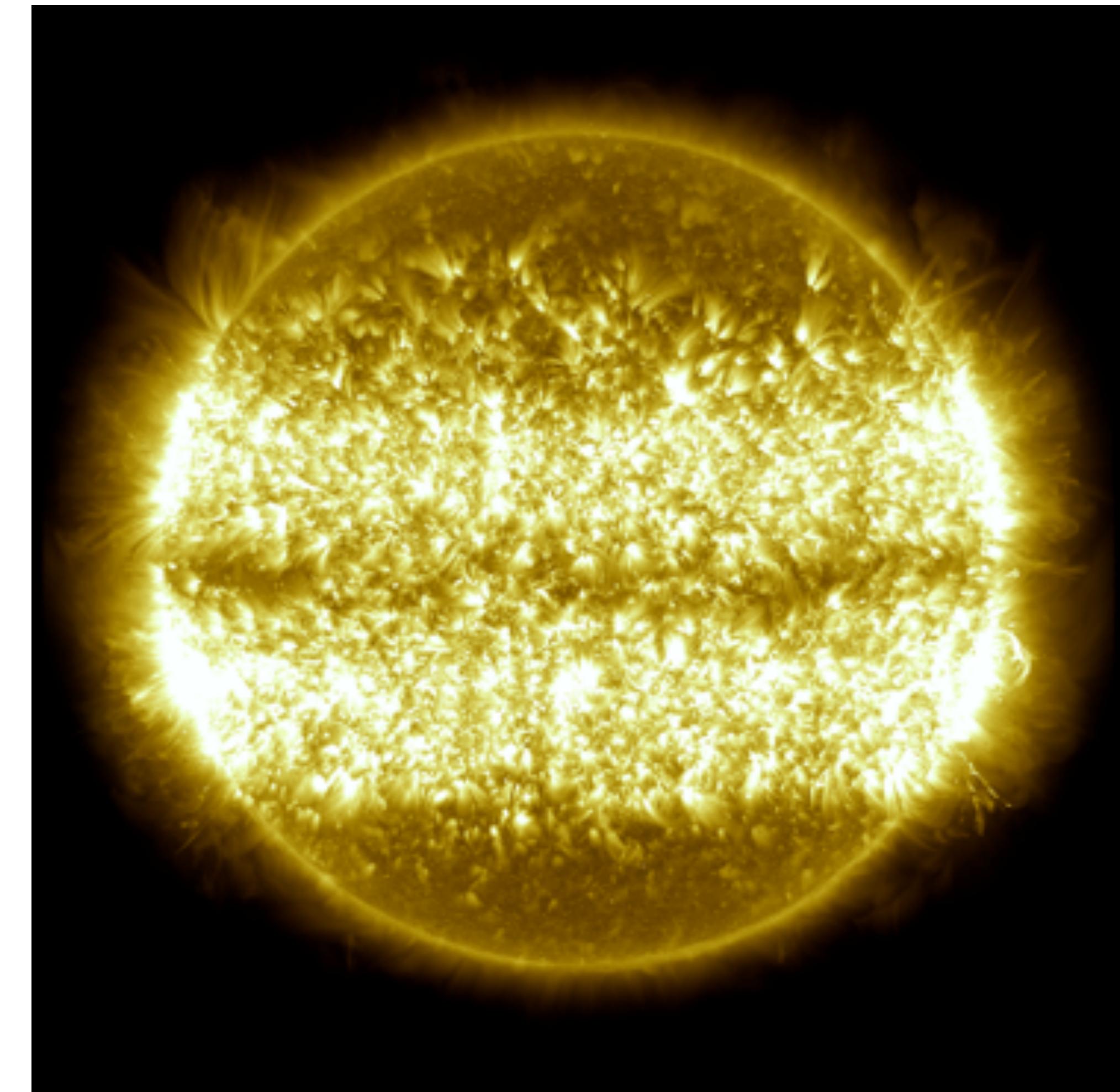
PDEs across science & engineering

Belousov–Zhabotinsky reaction



(Stephen Morris)

Solar convection



(NASA/SDO)

Scientific solvers: fast but narrowly focused

Design goals:

- Realistic parameters
- Maximum performance

Common limitations:

- Low-order methods
- Hard-coded models
- Difficult to modify



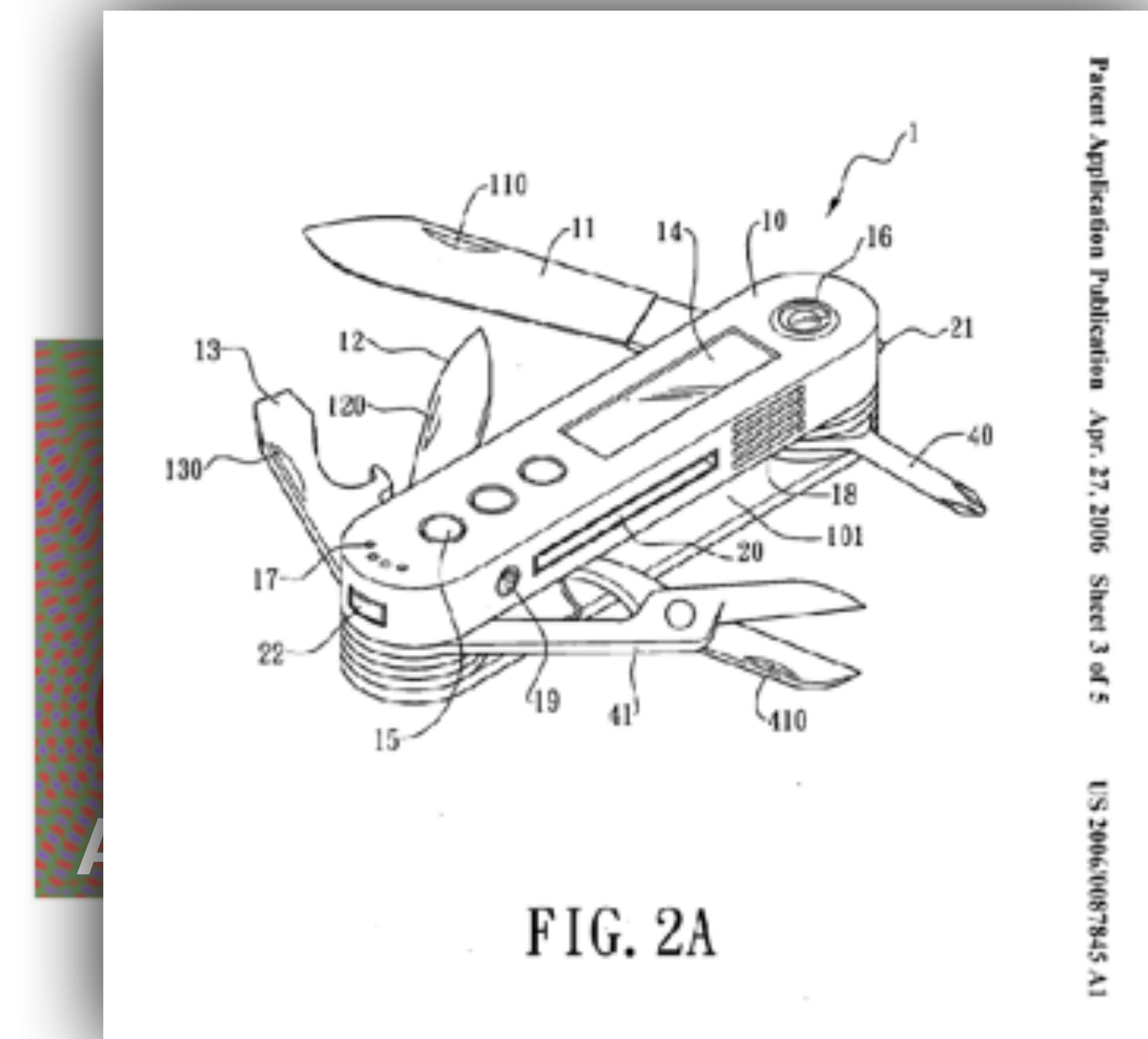
Mathematical solvers: flexible but slow

Design goals:

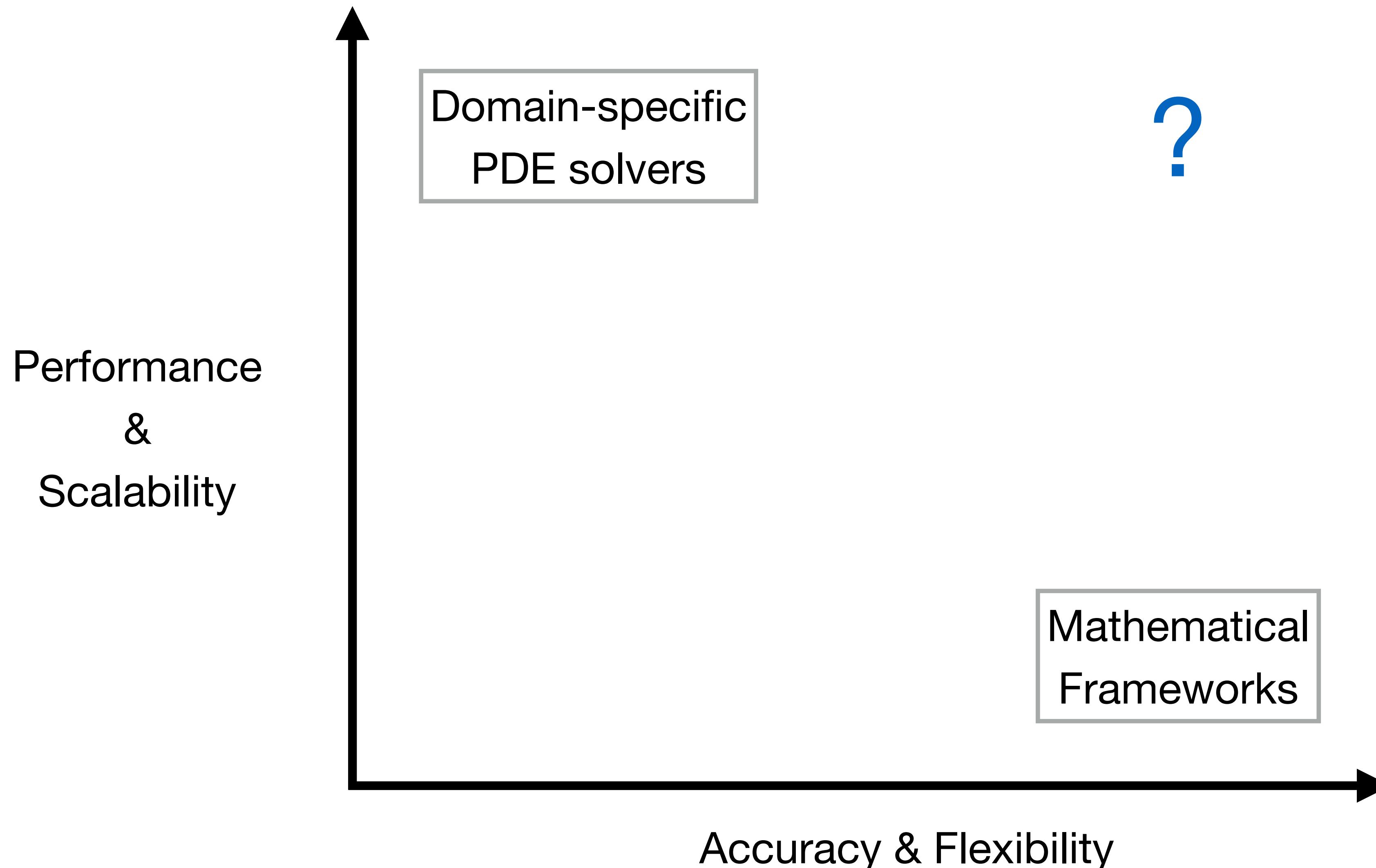
- Newest methods
- High accuracy

Common limitations:

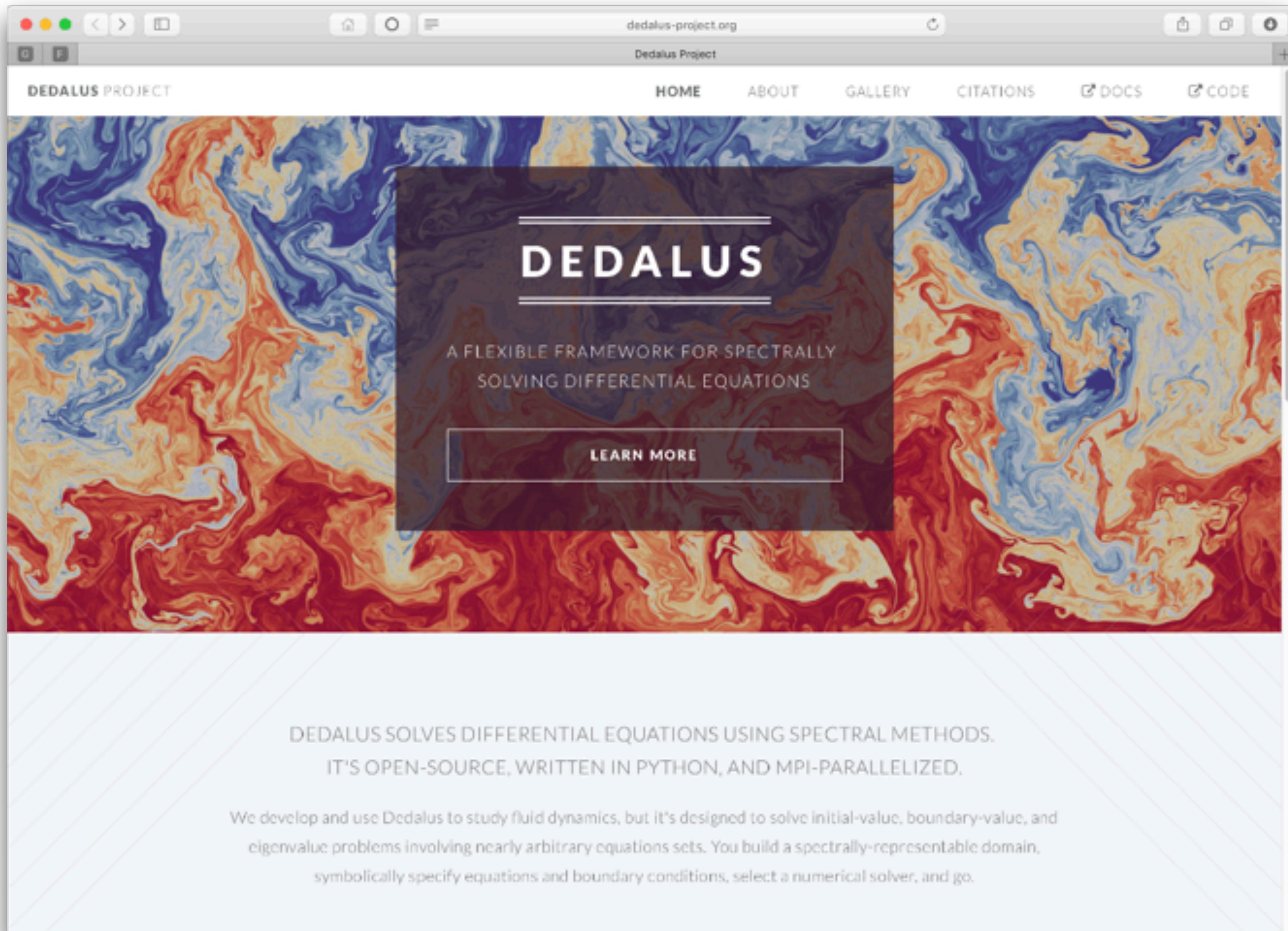
- Nonlinearly unstable
- Only scalar-valued fields
- Don't parallelize / scale



Challenge: high-order & flexible methods at scale



Dedalus Project



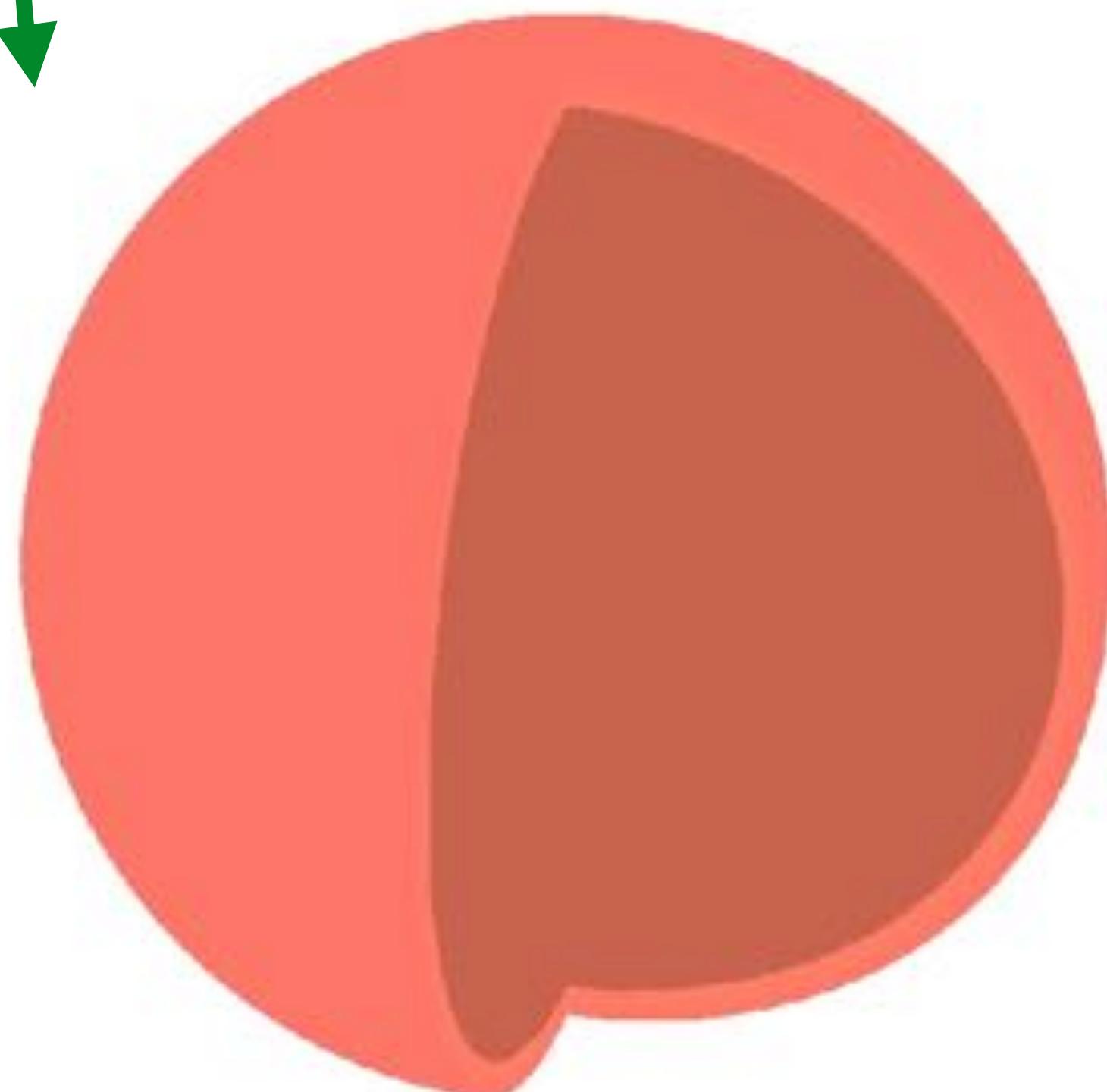
DEDALUS SOLVES DIFFERENTIAL EQUATIONS USING SPECTRAL METHODS.
IT'S OPEN-SOURCE, WRITTEN IN PYTHON, AND MPI-PARALLELIZED.

We develop and use Dedalus to study fluid dynamics, but it's designed to solve initial-value, boundary-value, and eigenvalue problems involving nearly arbitrary equations sets. You build a spectrally-representable domain, symbolically specify equations and boundary conditions, select a numerical solver, and go.

Dedalus Project

```
problem.add_equation("div(u) = 0")
problem.add_equation("dt(u) - v*Lap(u) + grad(p) + b*g = - u@grad(u)")
problem.add_equation("dt(b) - K*Lap(b) = - u@grad(b)")
```

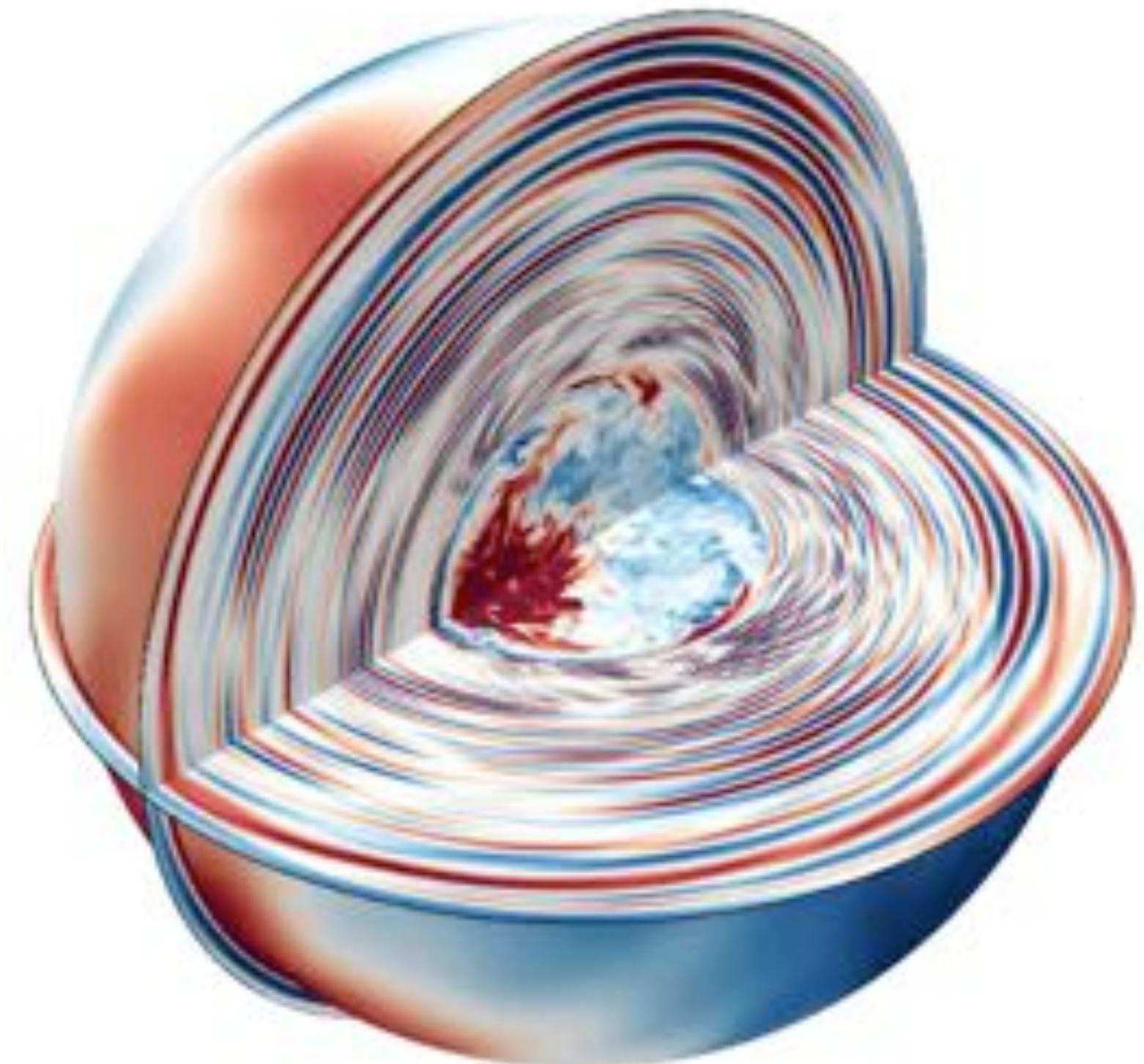
Rapid solver development
Spiral-defect chaos



Flexible equations
NLS quantum graphs



High performance
Turbulent wave excitation



Global Spectral Methods

Global spectral discretizations

Expand over “**trial**” functions:

$$u(x) = \sum_{n=0}^N u_n \phi_n(x)$$

Project equations against “**test**” functions:

$$\mathcal{L}u(x) = f(x)$$

$$\langle \psi_i | \mathcal{L}u \rangle = \langle \psi_i | f \rangle$$

$$\sum_j \langle \psi_i | \mathcal{L}\phi_j \rangle u_j = \langle \psi_i | f \rangle$$

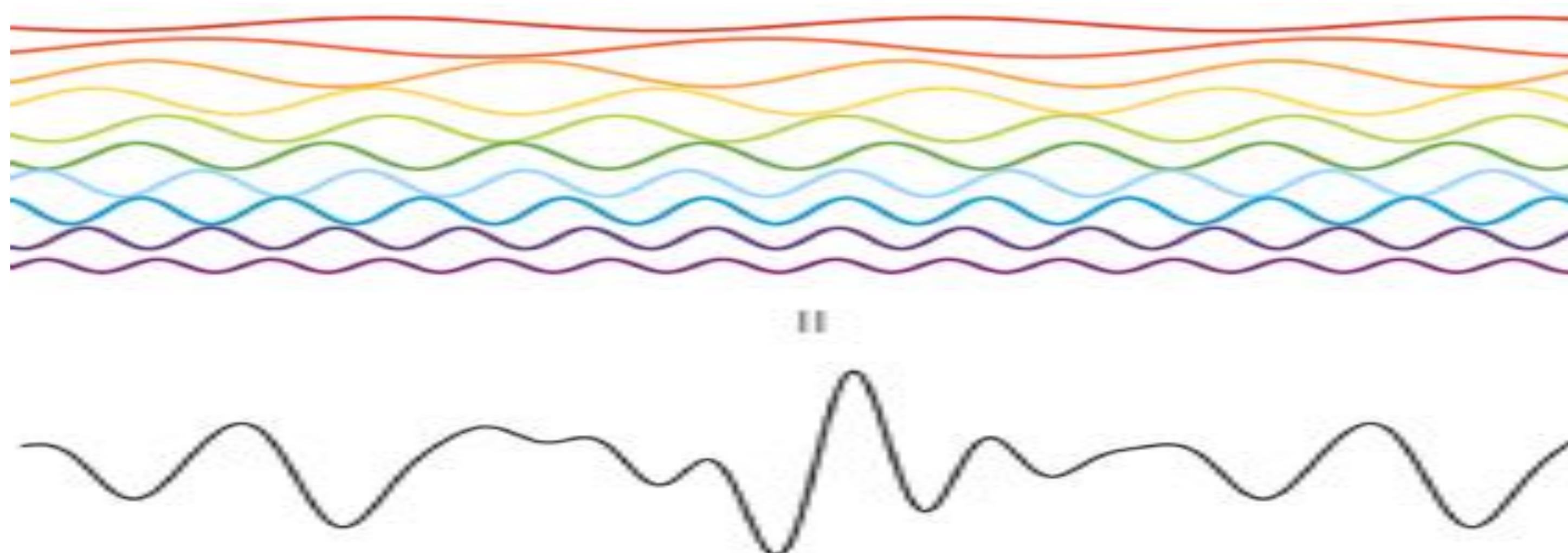
- Easy to adapt to different equations
- **Exponential convergence** for smooth functions
- Only possible in **simple geometries**
- **Fast** if discretized operators are **sparse**
- RHS terms require **spectral transforms**

Fourier spectral methods

Fourier series $\phi_n(x) = e^{inx}$

- **Exponential convergence** for smooth **periodic** functions
- **Fast transforms** for computing coefficients
- **Diagonal** derivative matrix:

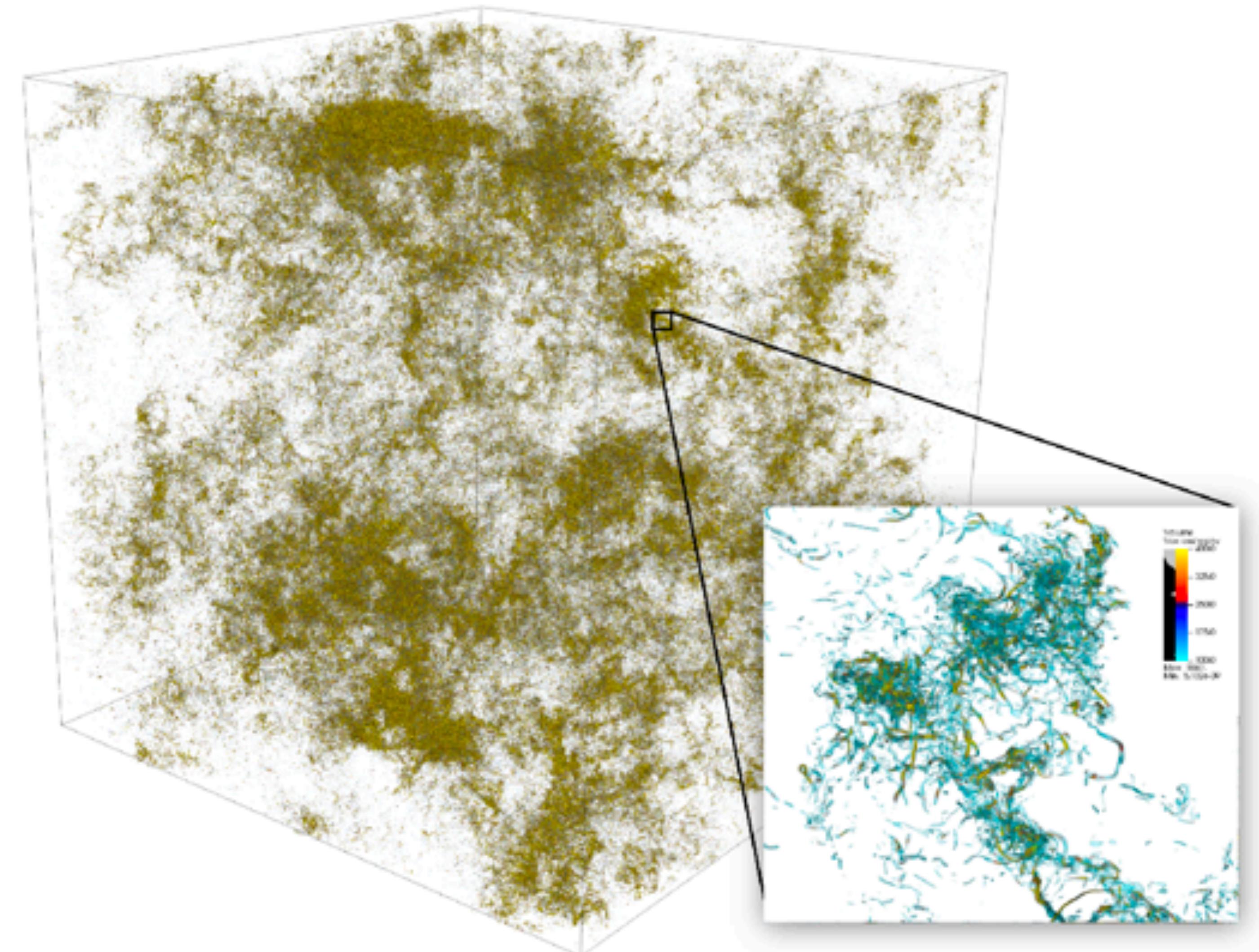
$$\langle \phi_m | \partial_x \phi_n \rangle = in\delta_{m,n}$$



World's largest turbulence simulations

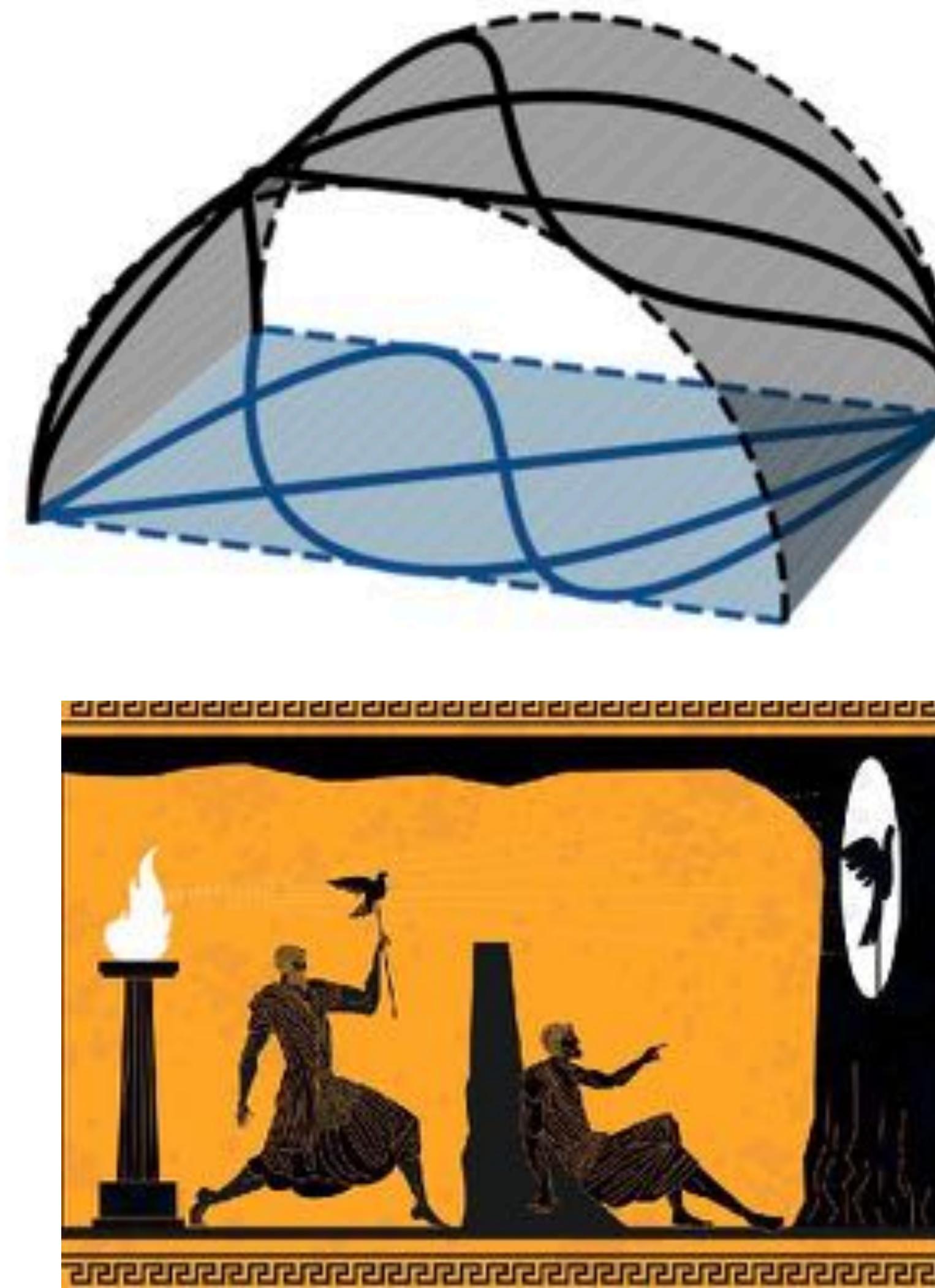
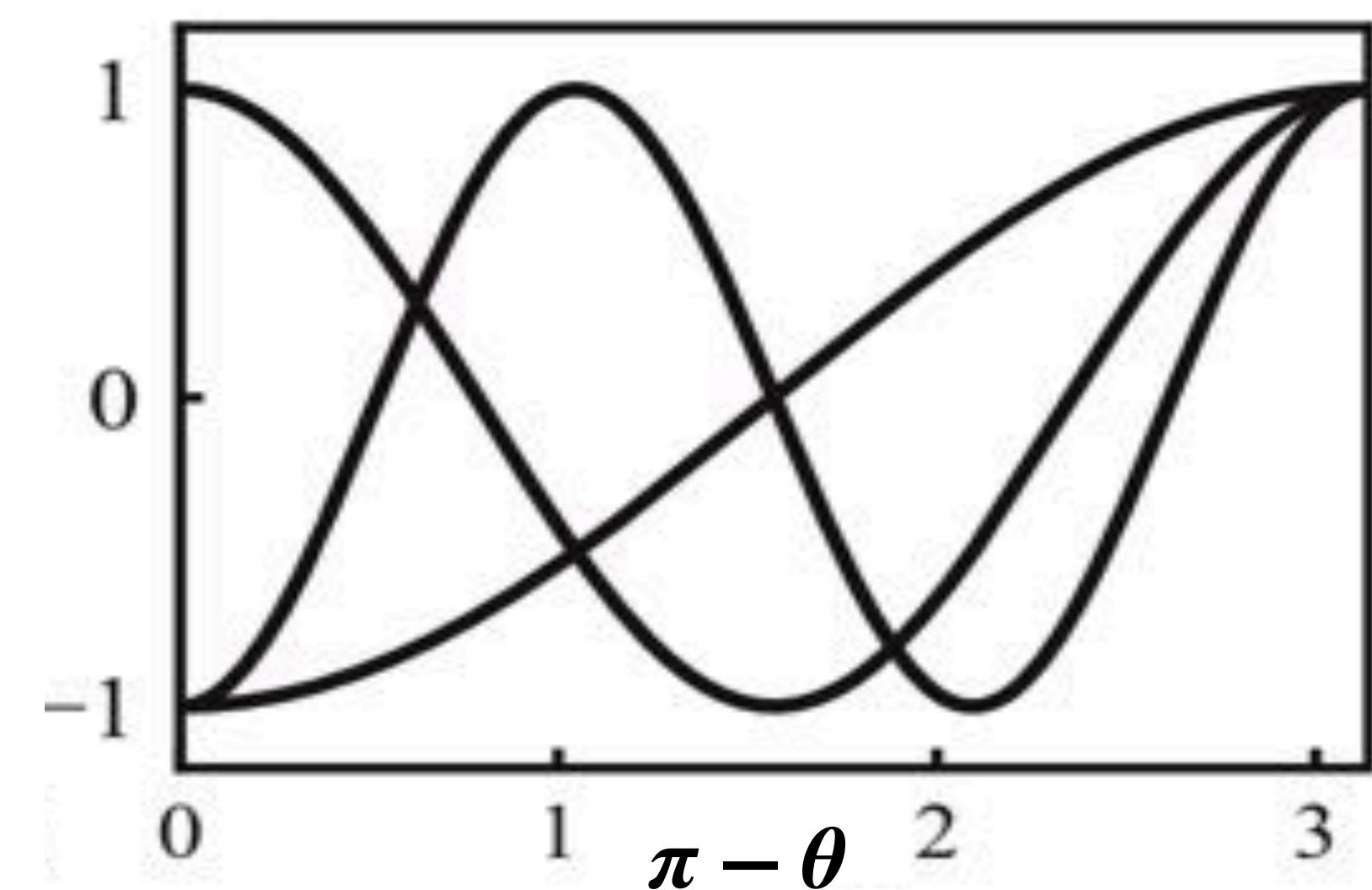
Yeung & Ravikumar, Phys. Rev. Fluids (2021)

- Fourier pseudospectral method (not Dedalus)
- $18,432^3$ grid points
- 18,432 GPUs

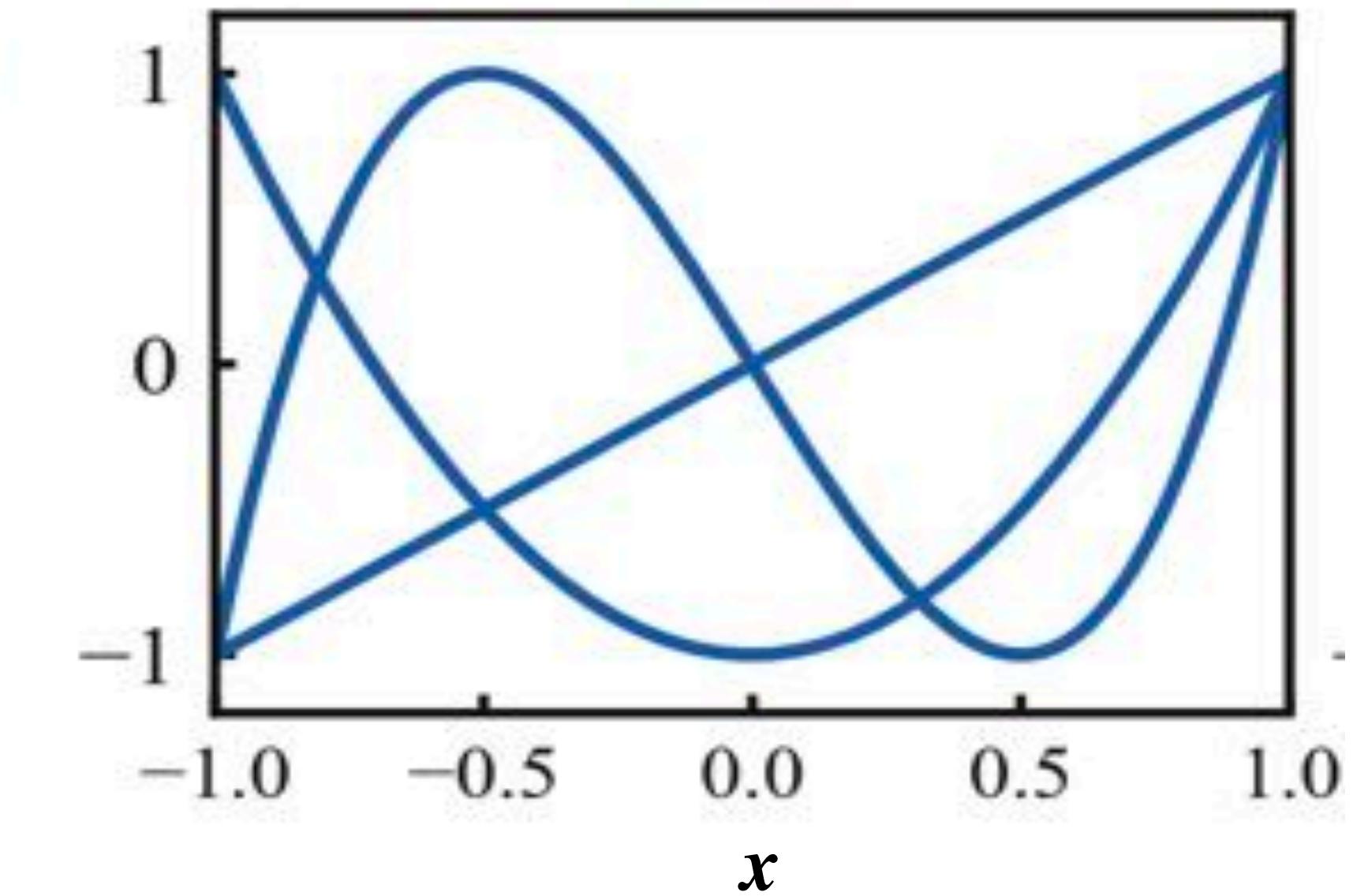


Chebyshev polynomials: cosines in disguise

$\cos(n\theta)$



$T_n(x)$



Orthogonal polynomials for non-periodic intervals

Jacobi polynomials $P_n^{(\alpha,\beta)}(x) \in \Pi_n$

- Orthogonal under weight: $w(x) = (1 - x)^\alpha(1 + x)^\beta$
- Closed under differentiation: $\partial_x^k P_n^{(\alpha,\beta)} \propto P_{n-k}^{(\alpha+k,\beta+k)}$
- **Exponential convergence** for smooth functions on $[-1, 1]$

1) **Legendre polynomials** ($\alpha = \beta = 0$) $P_n(x)$

- **Best L2 approximations** $w(x) = 1$

2) **Chebyshev polynomials** ($\alpha = \beta = -1/2$) $T_n(x)$

- **Fast transforms** (DCT) for computing coefficients

3) **Ultraspherical / Gegenbauer polynomials** ($\alpha = \beta = k - 1/2$) $C_n^{(k)}(x)$

- k -th derivatives of Chebyshev polynomials

Classical Chebyshev Methods

Same trial & test functions:

E.g. Legendre-tau

$$u(x) = \sum_{n=0}^N u_n P_n(x)$$

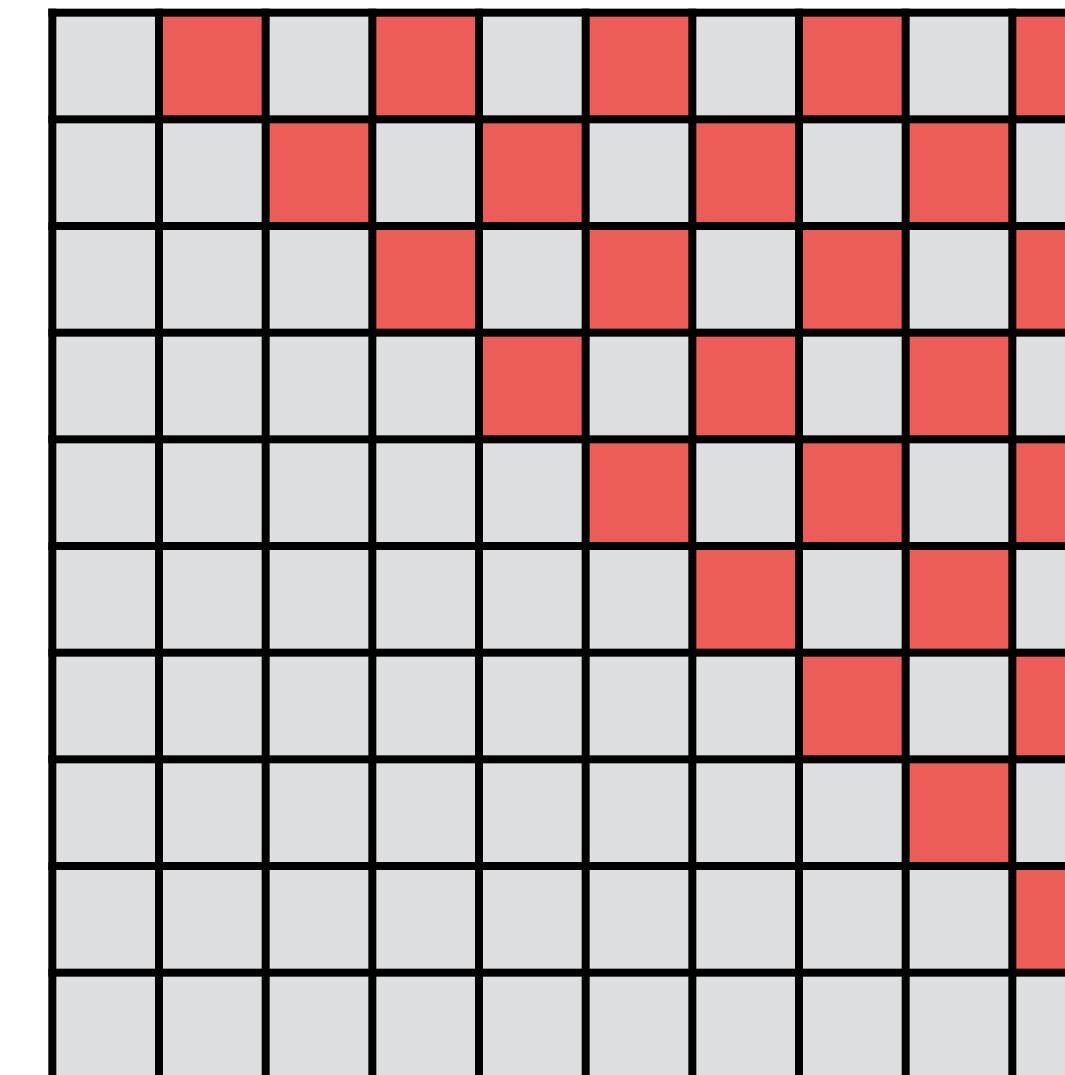
E.g. Chebyshev-tau

$$u(x) = \sum_{n=0}^N u_n T_n(x)$$

$$\mathbf{u}_n = \text{DCT}(u(x_i))$$

Differentiation:

$$\mathcal{D}_{m,n} = \langle T_m | \partial_x T_n \rangle$$



- Dense matrices
- Poor conditioning

Ultraspherical Method

Chebyshev trial functions:

$$u(x) = \sum_{n=0}^N u_n T_n(x)$$

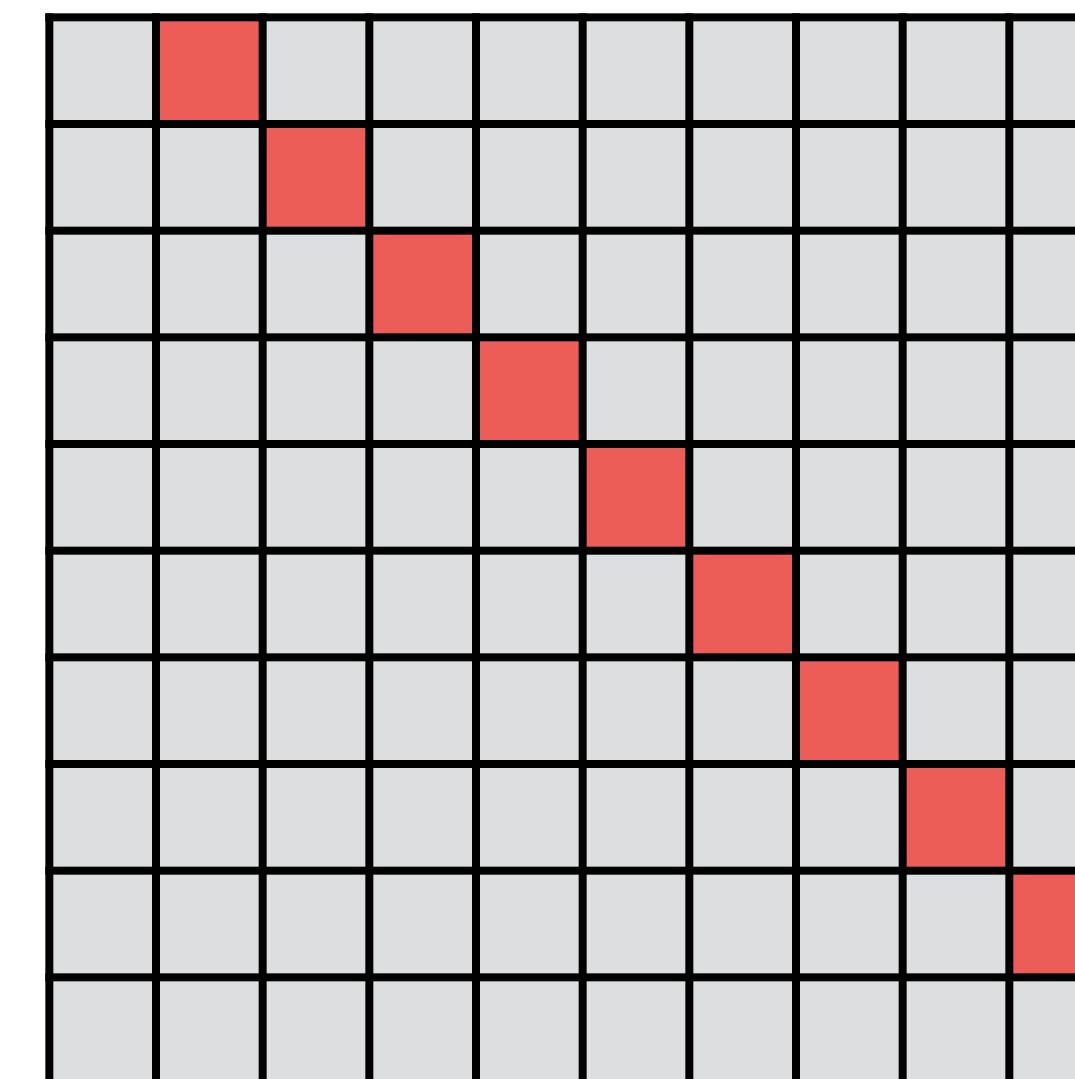
Ultraspherical test functions:

$$\alpha = \beta = k - 1/2$$

$$C_n^{(k)}(x) \propto \partial_x^k T_{n+k}(x)$$

Differentiation:

$$\mathcal{D}_{m,n} = \langle C_m^{(1)} | \partial_x T_n \rangle$$



- Banded
- Well conditioned

Key points for efficient spectral solvers

1. Spectrally accurate bases

- Rapidly convergent approximations

$$\{\phi_i(x)\}$$

2. Sparse differential operators

- Fast operator evaluation
- Fast direct solvers for LHS

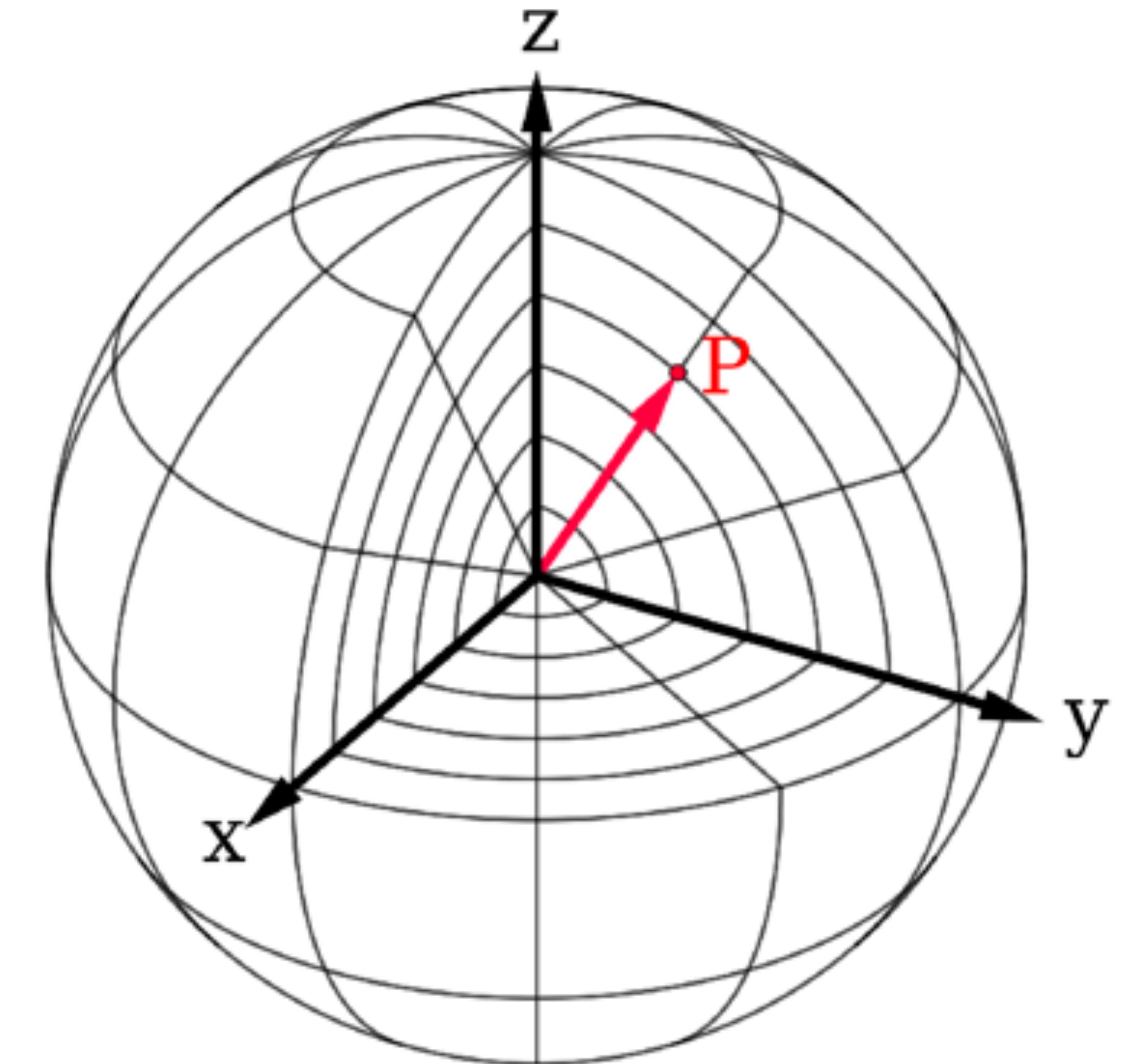
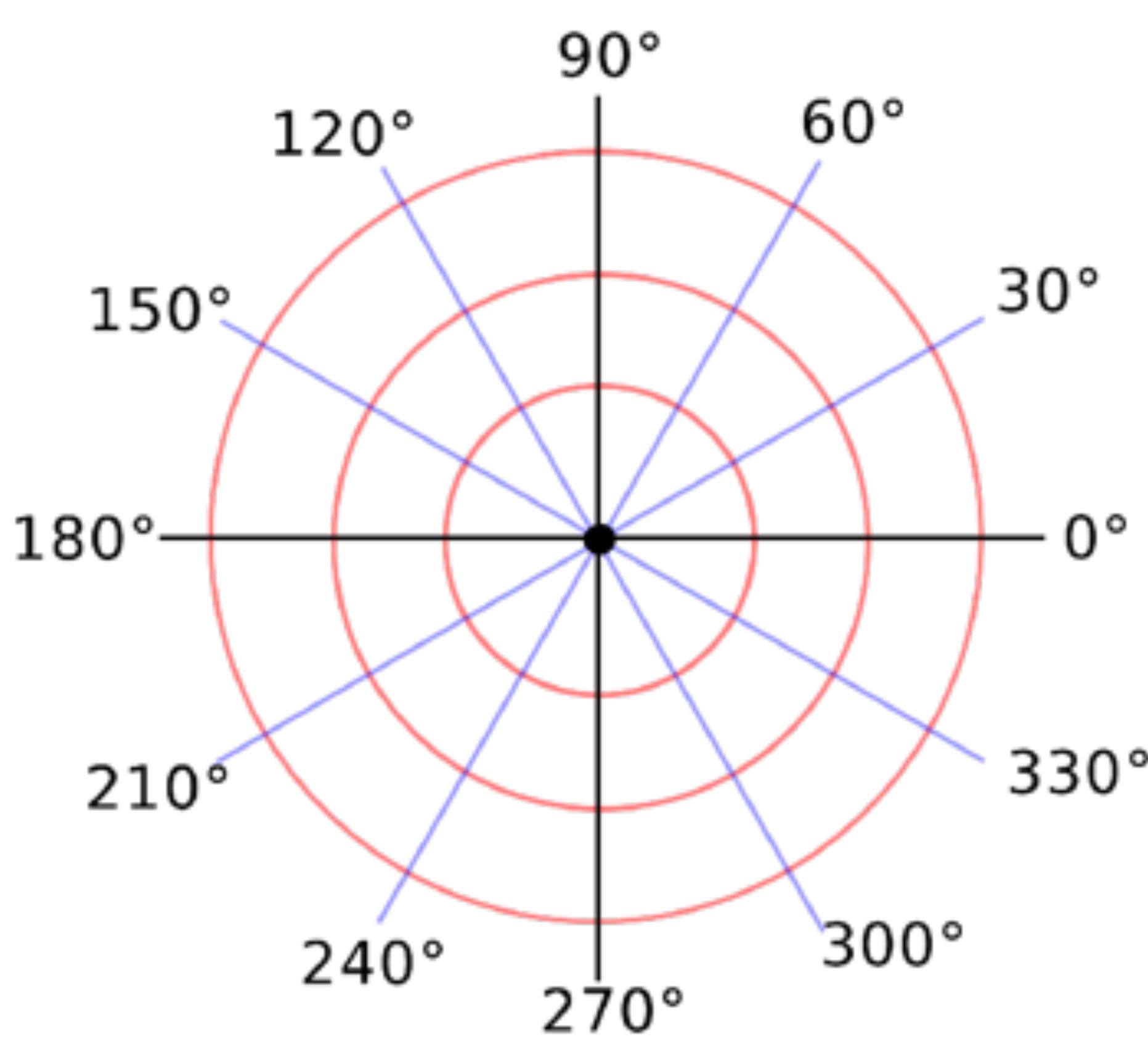
$$\langle \psi_i | H \phi_j \rangle$$

3. Fast spectral transforms

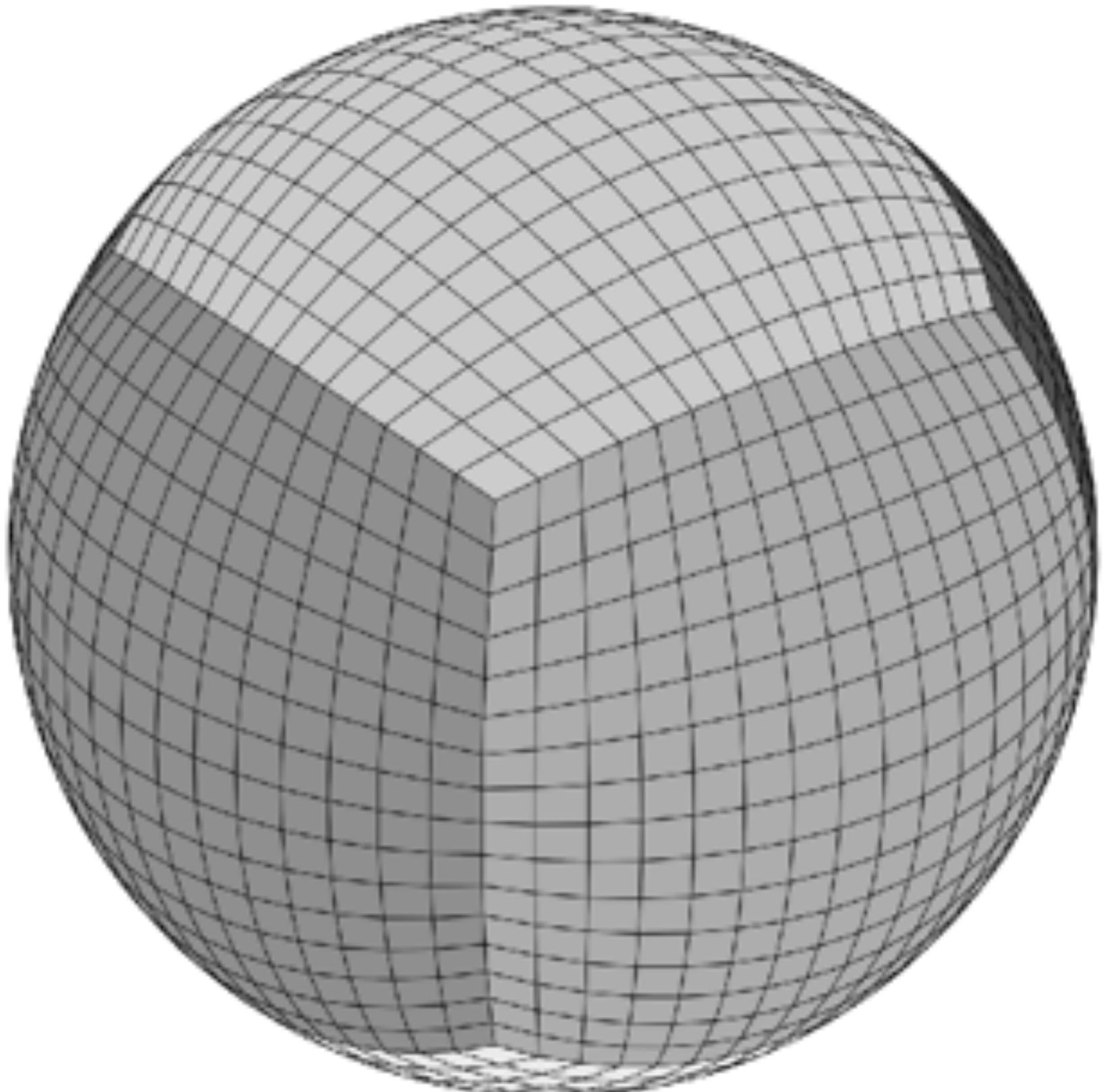
- Fast evaluation of nonlinear RHS

$$F(X)$$

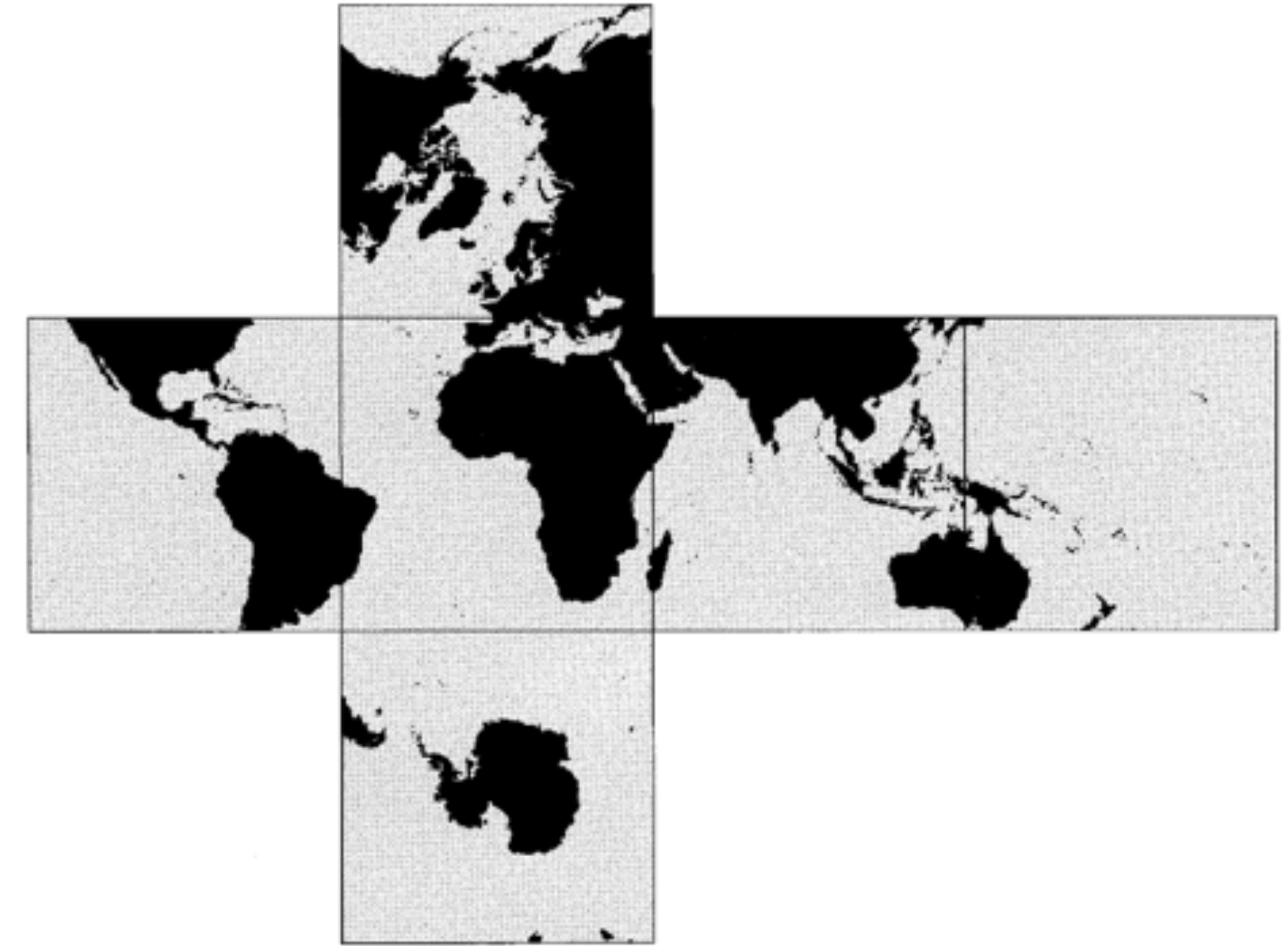
Polar & spherical coordinate singularities



The “cubed sphere” avoids the poles



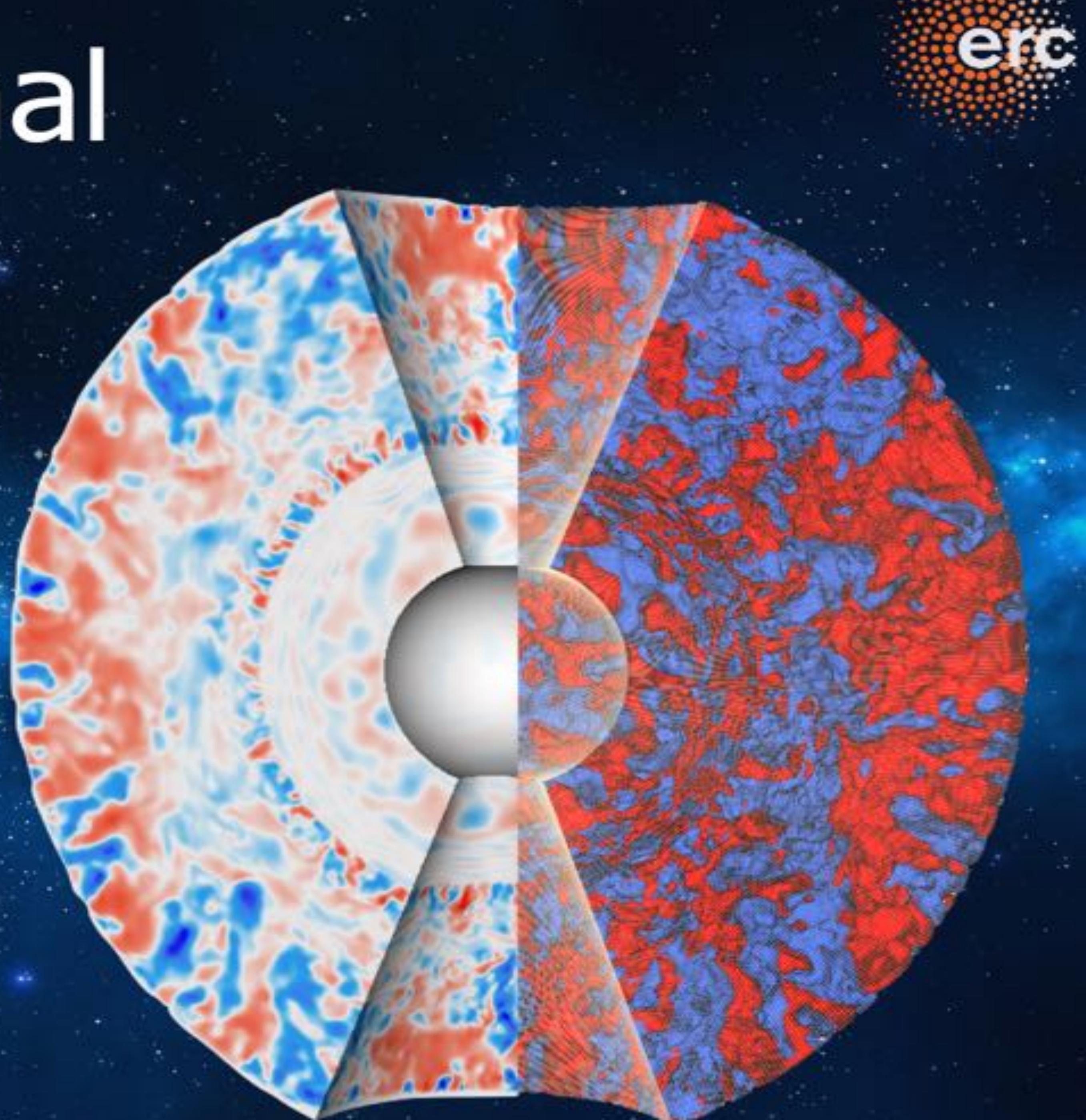
Ullrich (2014)



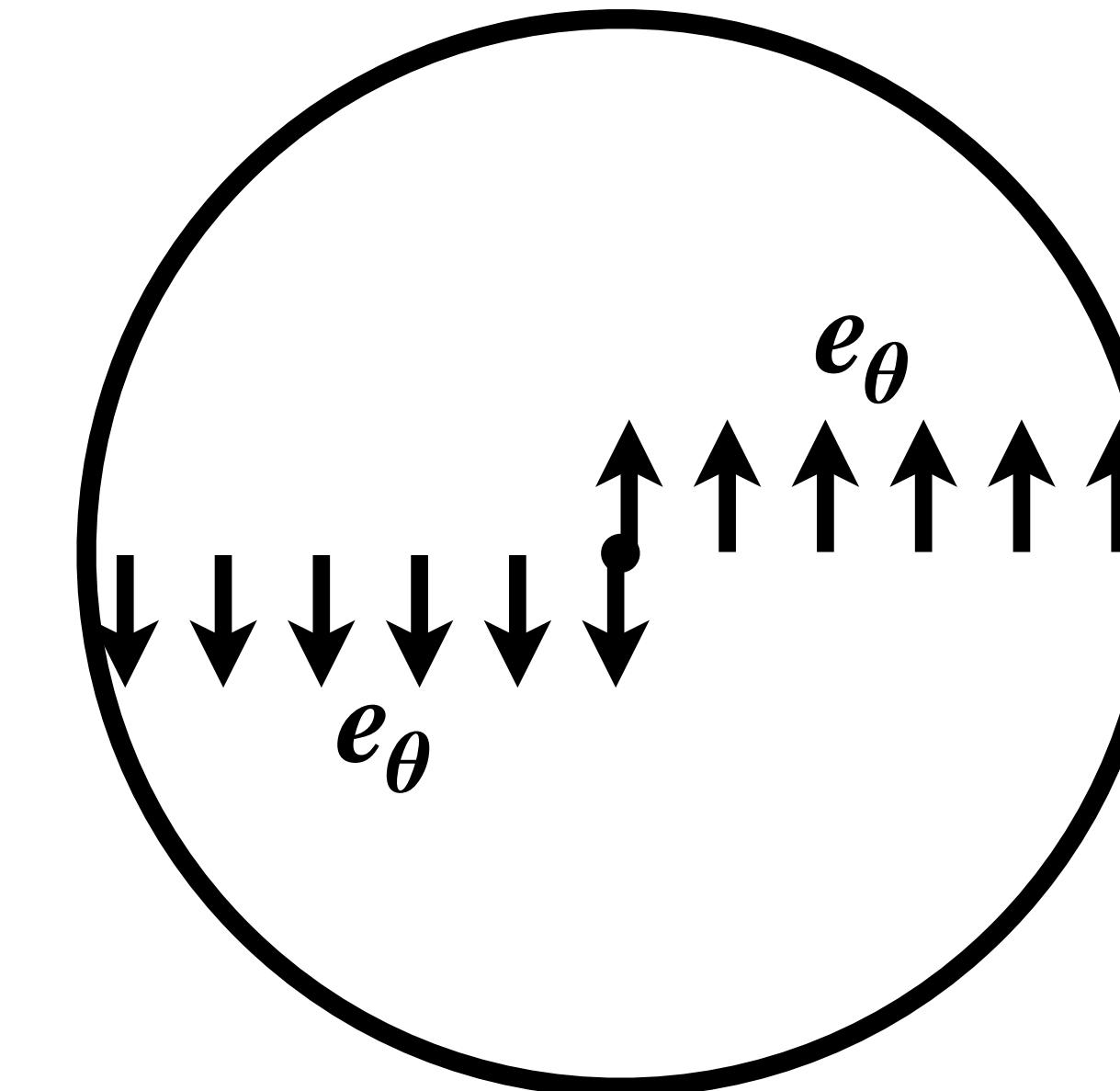
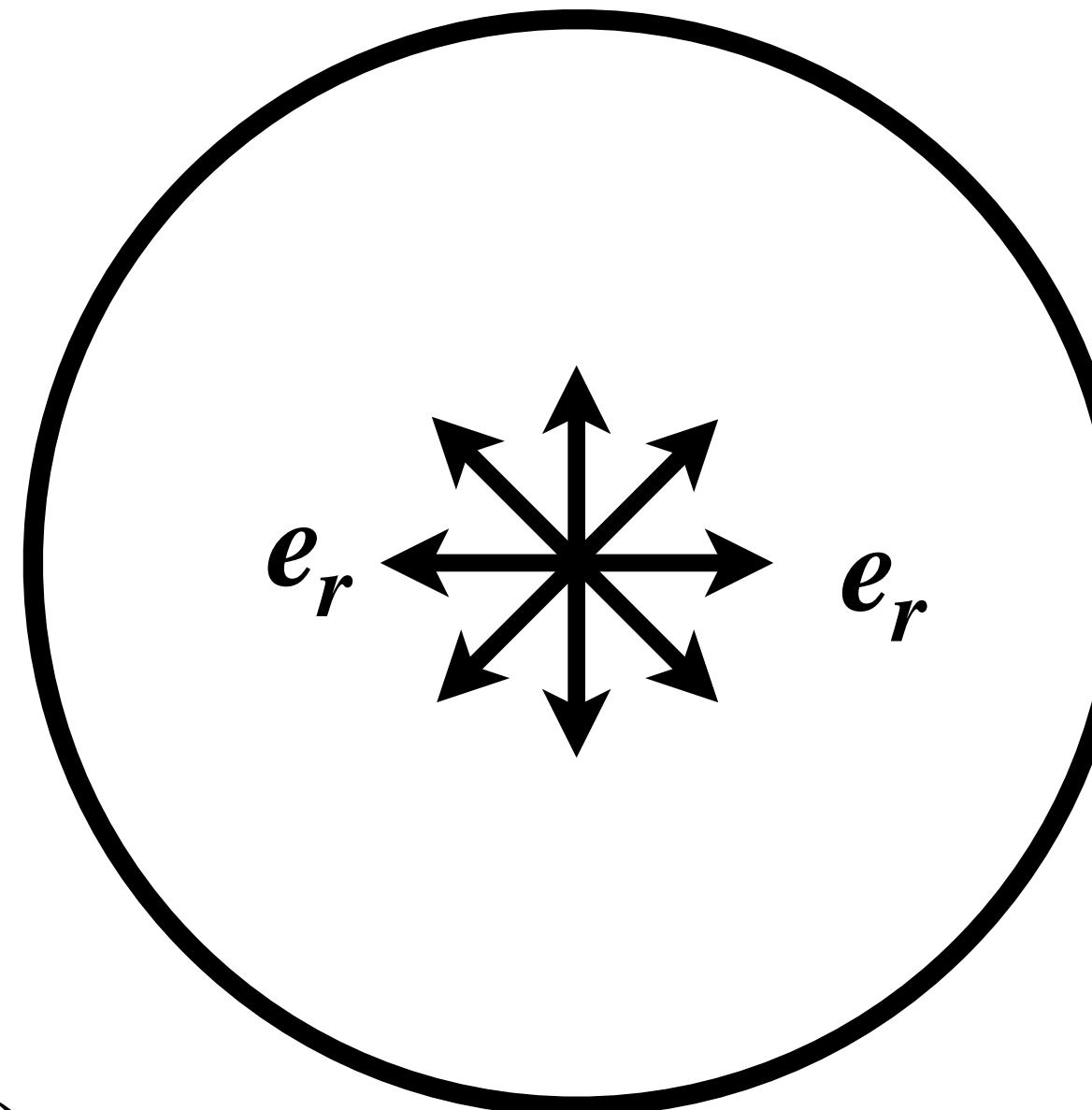
Rochi (1996)

Most codes also cut out the origin

MULTIDIMENSIONAL Stellar Implicit Code



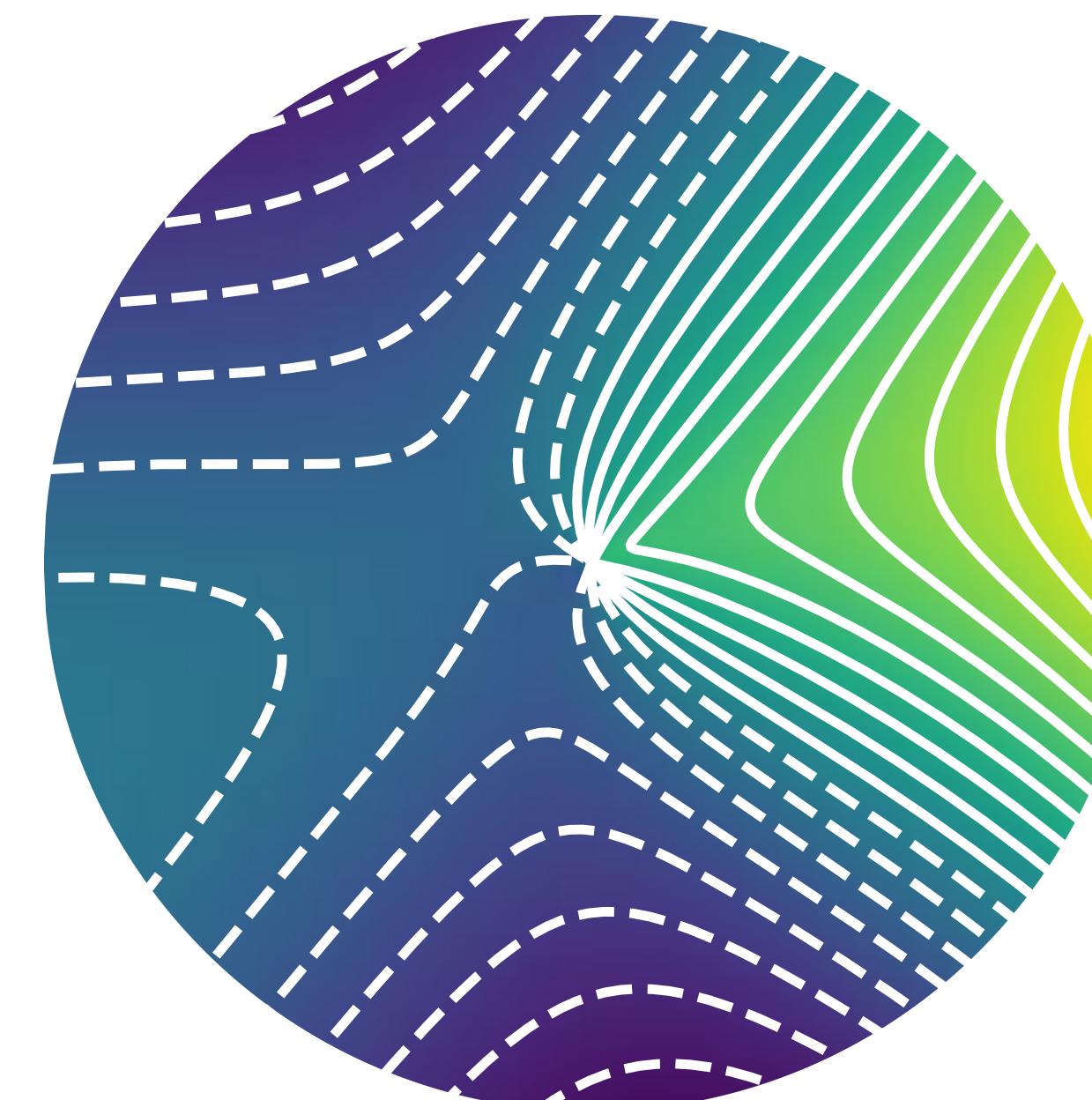
Components of smooth tensors become singular



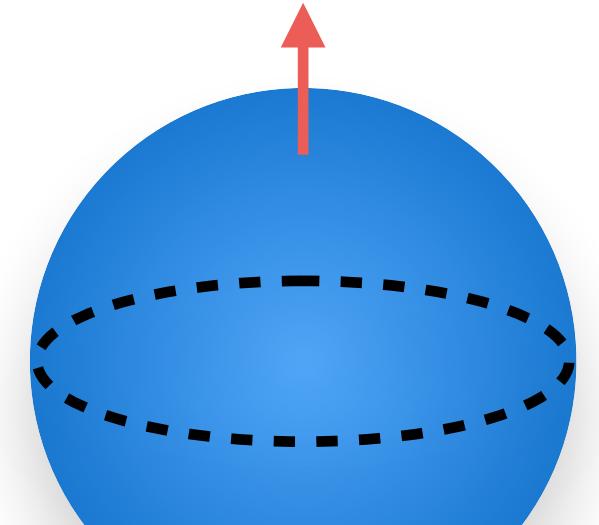
$f(x)$

$e_r \cdot \nabla f$

$e_\theta \cdot \nabla f$

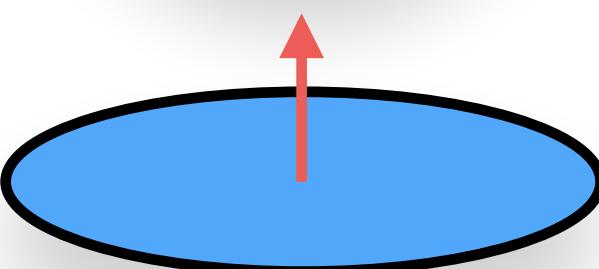


Regularity-aware curvilinear trial functions



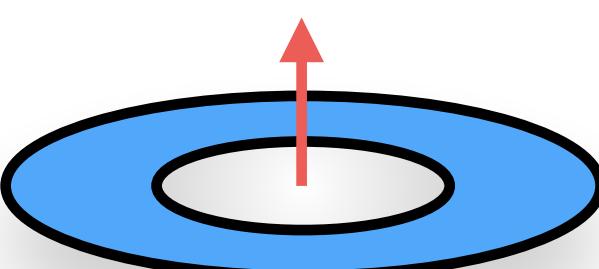
Spheres*: spin-weighted spherical harmonics
Newman & Penrose, JMP (1966)

$$Y_{l,m}^s(\phi, \theta)$$



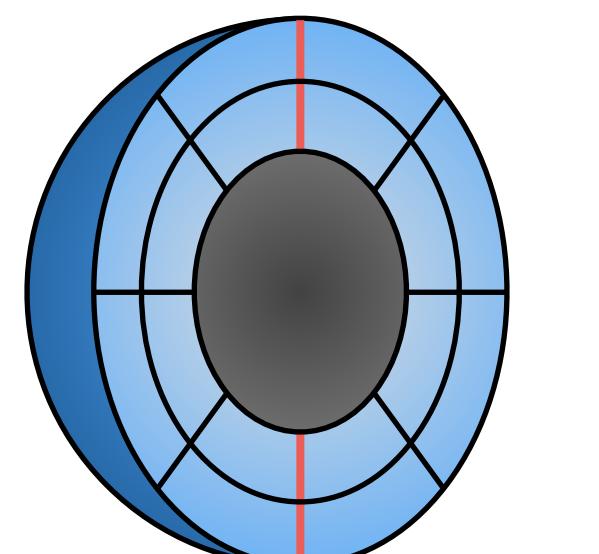
Disks: Fourier + generalized Zernike polynomials
Vasil et al (+KB), JCP (2016)

$$e^{im\phi} r^{m+s} P_n^{(k,m+s)}(r')$$



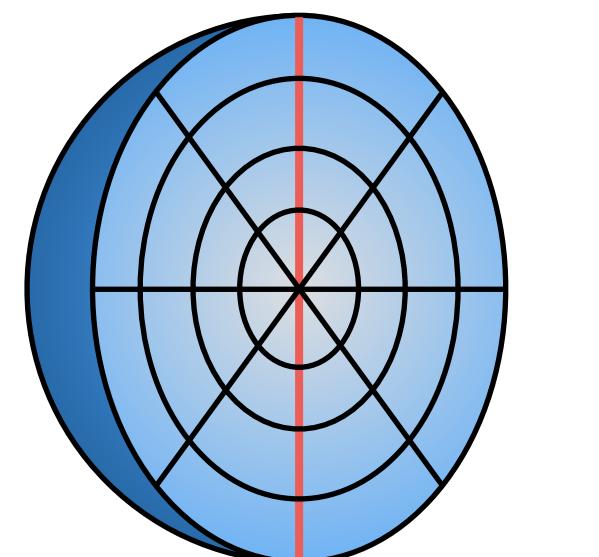
Arbitrary domains
Spectral accuracy for arbitrary tensor fields
Sparse tensor calculus

$$e^{im\phi} r^{-k} T_n(r')$$



Spherical shells: SWSH + rational Chebyshev
Dedalus collab (+KB), in prep.

$$Y_{l,m}^s r^{-k} T_n(r')$$



Balls: SWSH + one-sided Jacobi polynomials
Vasil et al (+KB), JCPX (2019)

$$Y_{l,m}^s Q_l^{s,a} r^{l+a} P_n^{(k,l+a+1/2)}(r')$$

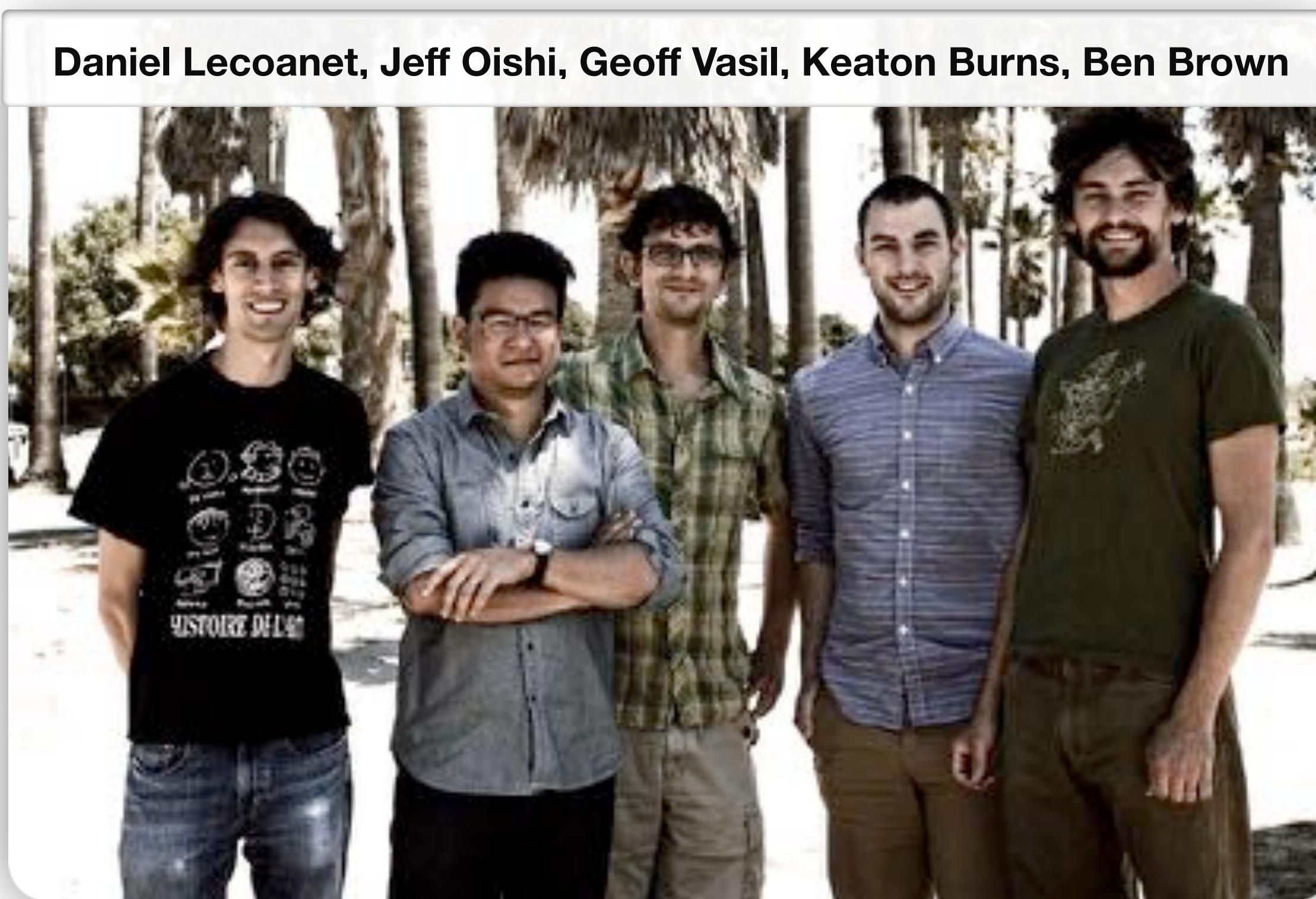
Dedalus Project

Dedalus Project

Community

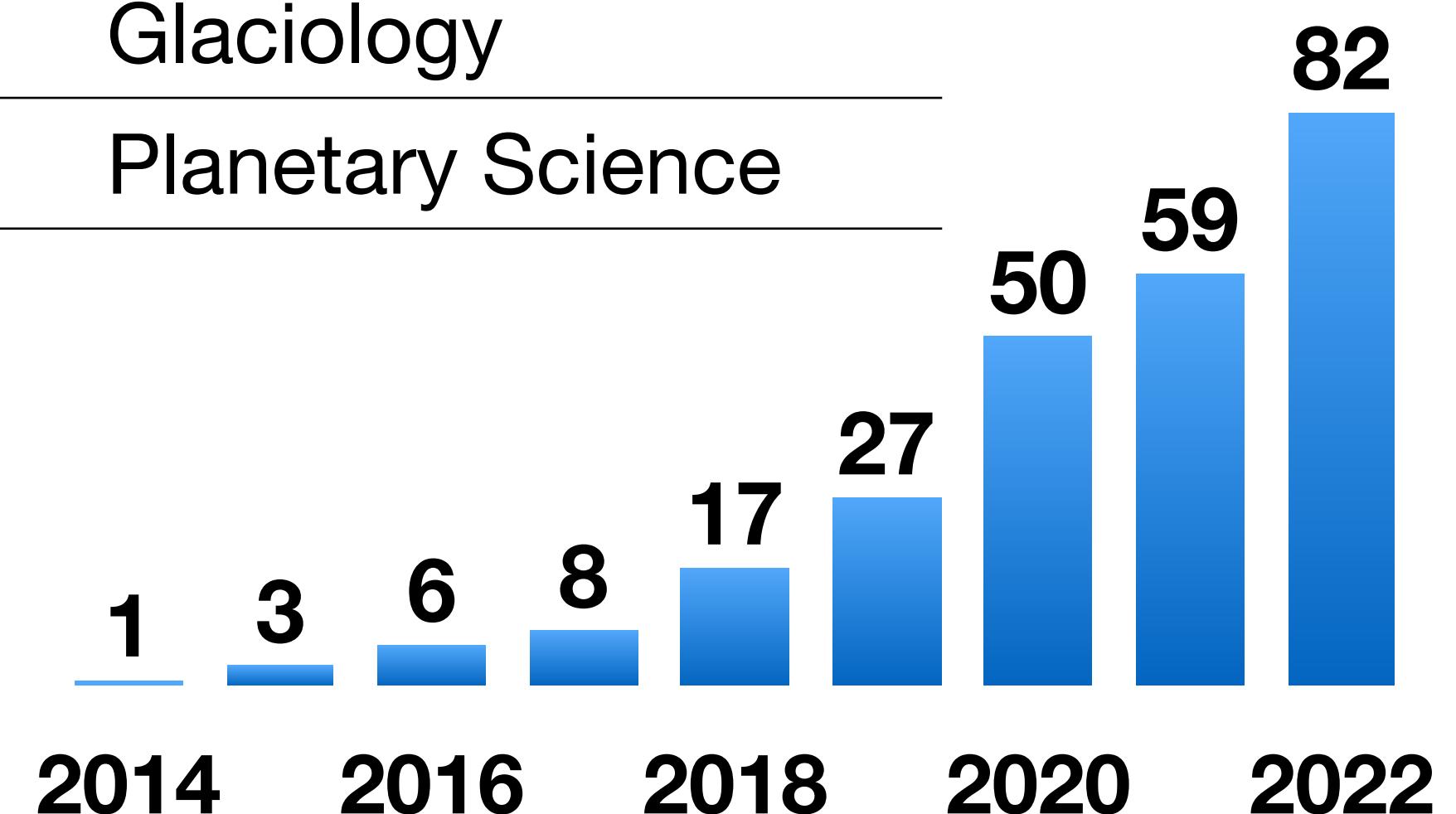
- 350+ members on user mailing list
- 19 contributors on GitHub
- NASA *High-value Open Source Tools* project

Core developers



Publications (250+)

25%	Fluid Dynamics
22%	Astrophysics
13%	Numerical analysis
10%	Plasma Physics
9%	Oceanography
6%	Atmospheric Science
5%	Biology
4%	Condensed Matter
3%	Glaciology
3%	Planetary Science



Supported problem types

Initial value problems:

$$\mathcal{M} \cdot \partial_t \mathcal{X} + \mathcal{L} \cdot \mathcal{X} = \mathcal{F}(\mathcal{X})$$

Nonlinear boundary value problems:

$$\mathcal{L} \cdot \mathcal{X} = \mathcal{F}(\mathcal{X})$$

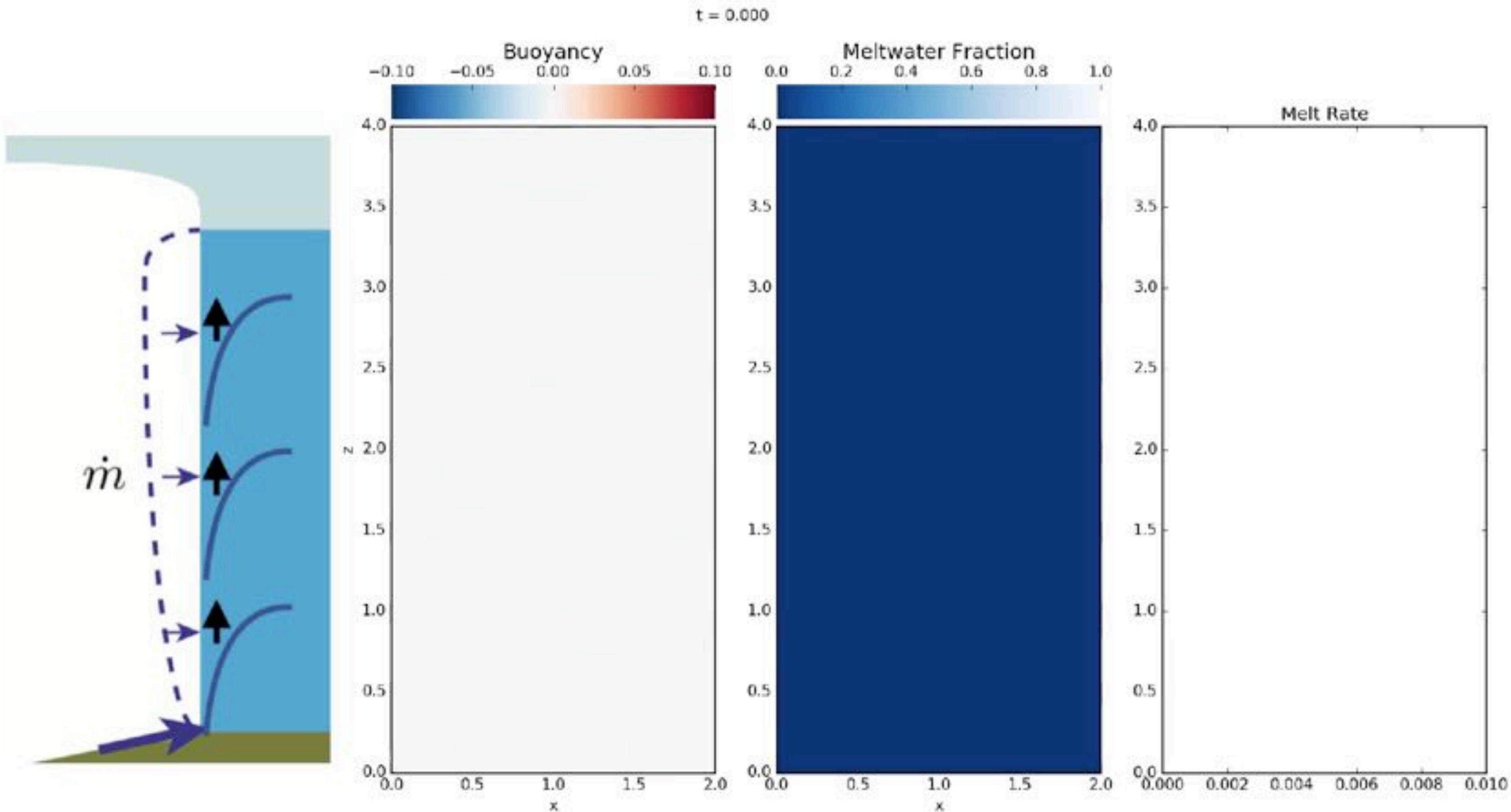
Eigenvalue problems:

$$\sigma \mathcal{M} \cdot \mathcal{X} + \mathcal{L} \cdot \mathcal{X} = 0$$

Pseudospectra:
(Eigentools package)

$$\sigma \mathcal{M} \cdot \mathcal{X} + (\mathcal{L} + \mathcal{N}) \cdot \mathcal{X} = 0, \quad \|\mathcal{N}\| \leq \epsilon$$

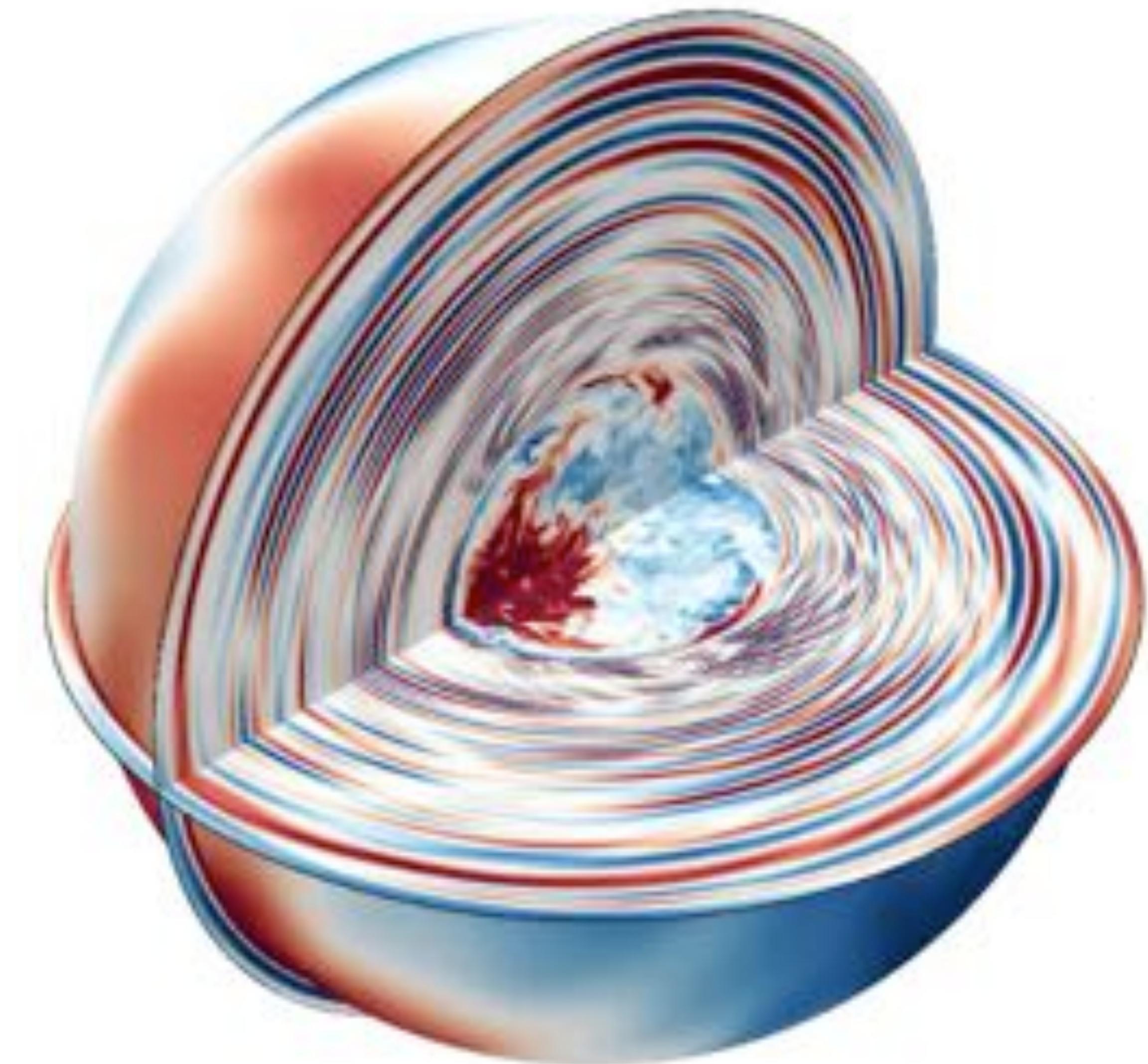
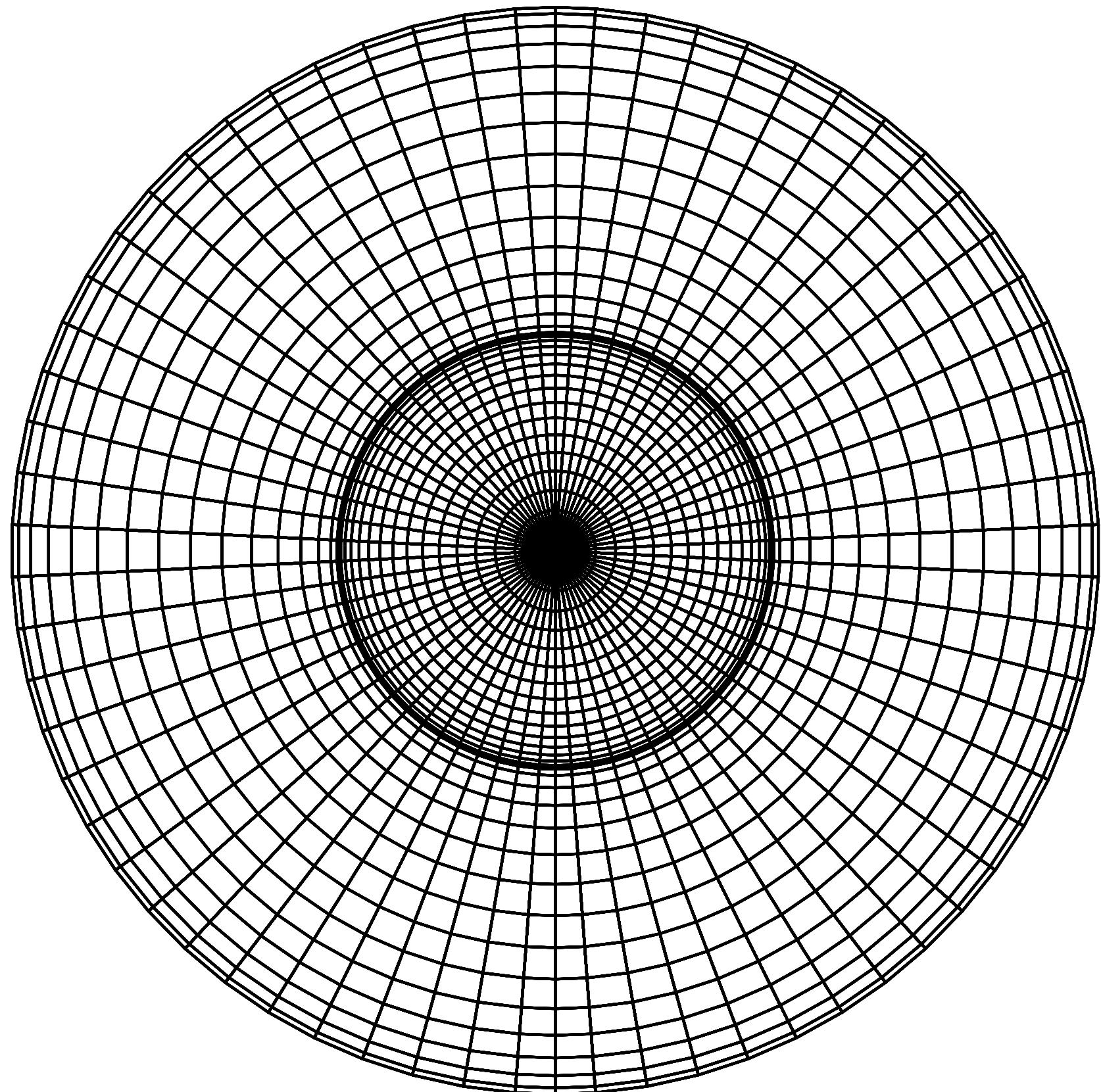
Turbulent enhancement of glacier melting



Burns (2018)

High- p spherical spectral elements

- Stacked ball and spherical shell bases
- Resolves internal/material boundaries



w/ Evan Anders

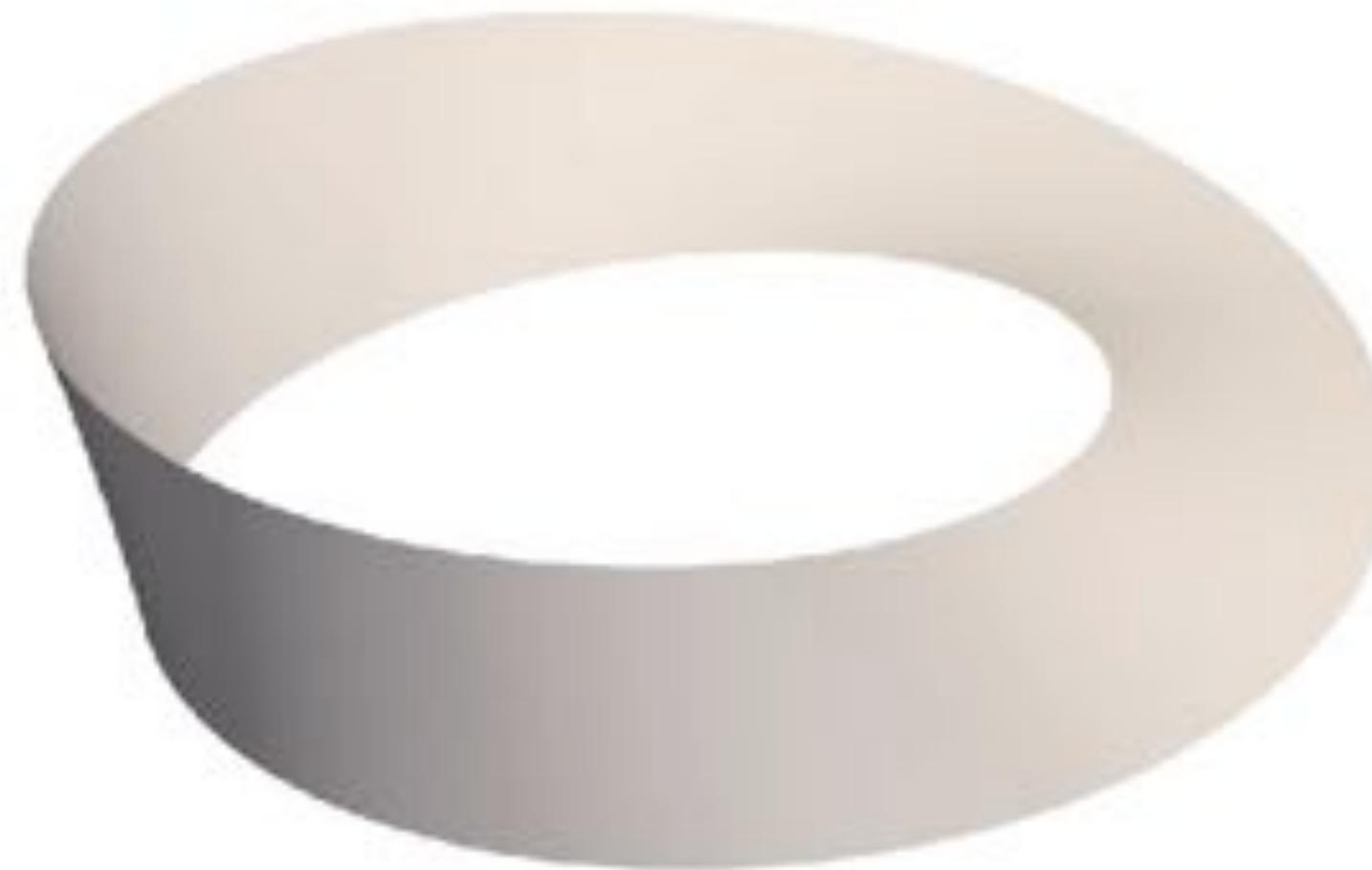
Quantum graphs

$$i\partial_t\psi + \frac{1}{2}\partial_x^2\psi = -|\psi|^2\psi$$



Non-orientable & symplectic manifolds

Möbius strip



Klein bottle



Real projective plane



Goals:

- Double-cover domains with exact symmetries
- Symplectic manifolds / phase-space simulations

