Tensor-product elliptic solver for liquid-metal magnetohydrodynamics

SuperMUC-NG Status and Results Workshop, May 9 – 11, 2023

Computing project: pn68ni



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Research network



TU Ilmenau⁽¹⁾





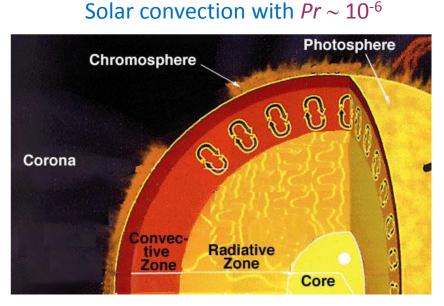


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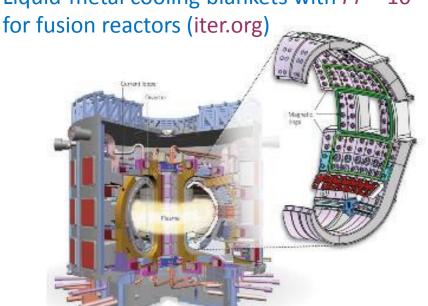
Research topics covered in 2022-2023

- 1. Turbulent convection at very low Prandtl numbers
- 2. Development of numerical tools for wall-bounded MHD flows
- **3**. Evolution of MHD flows in ducts and rectangular boxes

Applications:



https://eng.libretexts.org/@go/page/5948?pdf



Liquid-metal cooling blankets with $Pr \sim 10^{-2}$

Tools applied

1. In-house flow solver for incompressible flows in rectangular geometries (TU Ilmenau, UMICH Dearborn)

*Key features: based on 2nd order finite-differences, conservative scheme, structured collocated grids, MPI + Open MP hybrid parallelization

2. NEK 5000 solver, open-source community driven code (TU Ilmenau, NYU, NYU Abu-Dhabi)

*Key features: spatial discretization with spectral elements, unstructured grids, multi-domain decomposition, MPI parallelization

3. Experimental facilities with liquid mercury (Moscow) *Solenoid magnet up to 1.7T, temperature probes, electric potential sensors

Physical model:

Governing equations:

$$\nabla \cdot \boldsymbol{u} = 0$$

$$\frac{\partial \boldsymbol{u}}{\partial t} + (\boldsymbol{u} \cdot \nabla)\boldsymbol{u} = -\nabla \mathbf{p} + \sqrt{\frac{Pr}{Ra}} [\nabla^2 \boldsymbol{u} + Ha^2(\boldsymbol{j} \times \boldsymbol{e_z})] + T\boldsymbol{e_z}$$

$$\frac{\partial T}{\partial t} + (\boldsymbol{u} \cdot \nabla)T = \sqrt{\frac{1}{RaPr}} \nabla^2 T$$

$$\boldsymbol{j} = -\nabla \varphi + (\boldsymbol{u} \times \boldsymbol{e_z})$$

$$\nabla^2 \varphi = \nabla \cdot (\boldsymbol{u} \times \boldsymbol{e_z})$$

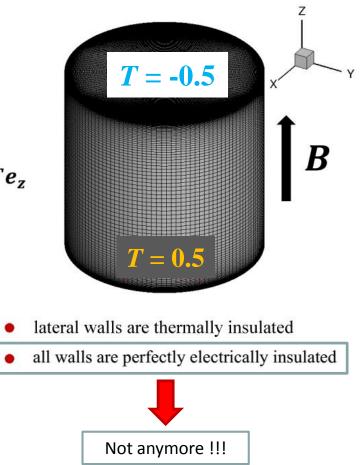
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> Approximations:

- The Boussinesq approximation
- The quasi-static model of electromagnetic interactions: $Re_m \ll 1$ and $Pr_m \ll 1$

Non-dimensional control parameters:

$$Pr = \frac{\nu}{\kappa}$$
 $Ra = \frac{g\alpha\Delta TH^3}{\nu\kappa}$ $Ha = B_o H \sqrt{\frac{\sigma}{\rho\nu}}$ $\Gamma = \frac{D}{H}$



#1: Turbulent convection at low Prandtl numbers

Parameters of simulations with FD in-house solver

rectangular box with aspect ratio L/H = 25/1 (width/height)

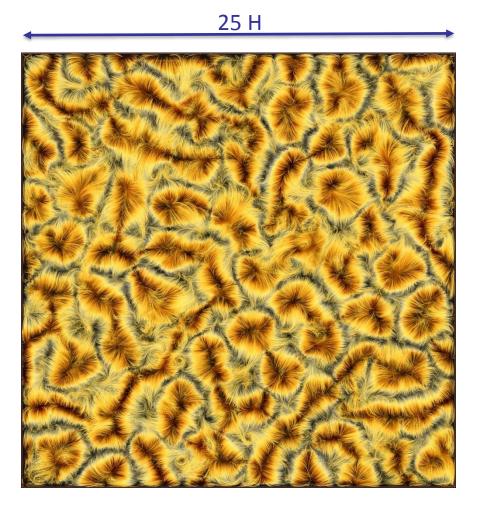
 $Ra = 10^5 \dots 10^7$ and Pr = 0.021, 0.005, 0.001

Simulation	Resolution	Computing information	Nu	Re
<i>Ra</i> = 1e6 , <i>Pr</i> = 0.021	8192 ² x 512	34 Bill. points, @ 24576 cores	4.74282523	3155
<i>Ra</i> = 1e6, <i>Pr</i> = 0.005	8192 ² x 512	34 Bill. points, @ 24576 cores	3.48979662	7638
<i>Ra</i> = 1e6, <i>Pr</i> = 0.001	12800 ² x 800	131 Bill. points, @ 38400 cores	2.47784979	19901
<i>Ra</i> = 1e5, <i>Pr</i> = 0.001	9600 ² x 640	60 Bill. points, @ 28800 cores	1.21289961	4776
<i>Ra</i> = 1e7, <i>Pr</i> = 0.001	20480 ² x 1280	0.54 Trill. points, @ 144000 cores	4.60336402	56103

Information about the largest simulation at $Ra = 10^7$ and $Pr = 10^{-3}$

- (*) consumed compute time totals 35 Mill. core-hours (one run at 144000 cores)
- (*) highest resolution maxed out at 22400² x 1400 = 0.7 Trill. points
- (*) one snapshot with 3D flow field exceeds 17 TB
- (*) total data to be post-processed more than 190 TB

#1: Turbulent convection at low Prandtl numbers Flow field at $Ra = 10^5$ and Pr = 0.001 and $\Gamma = 25$ (L/H)

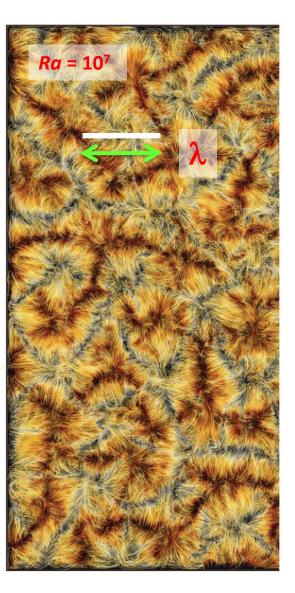


Superstructures shown by streamlines, seeded with 15000 lines

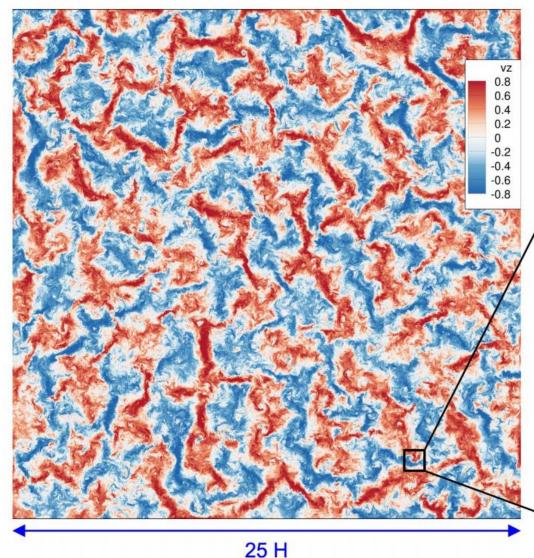
#1: Turbulent convection at low Prandtl numbers Flow fields at Pr = 0.001







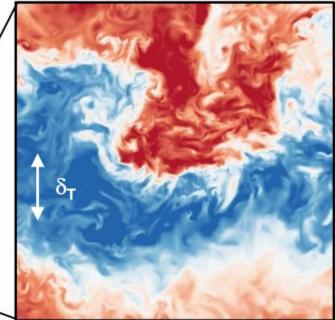
#1: Turbulent convection at low Prandtl numbers Vertical velocity Vz in the mid-plane at $Ra = 10^6$ and Pr = 0.001



On the left – entire domain, visualized at full resolution of 12800 × 12800 pixels (i.e. one pixel = one grid point)

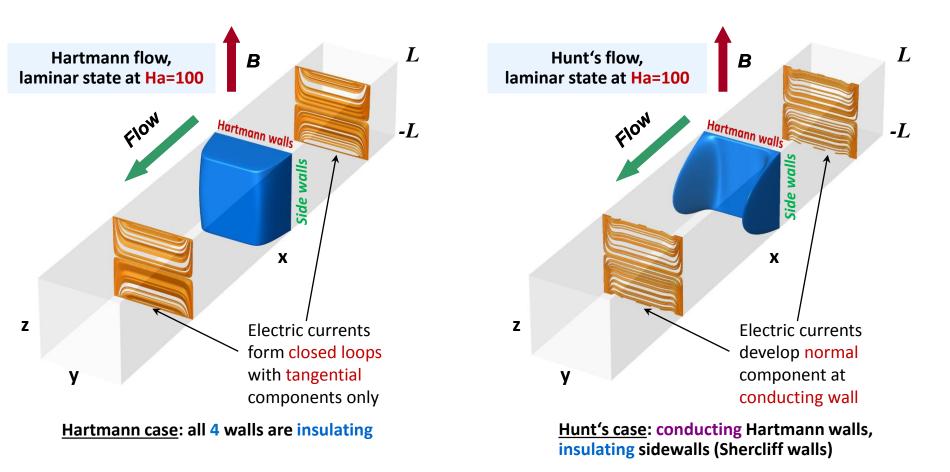
Zoom into a 1/1 region, displaying highly intertial small-scale turbulence

 δ_{T} indicates thickness of the thermal boundary layer



Important finding: Turbulence in the bulk is Kolmogorov-like with $E(k) \sim k^{-5/3}$

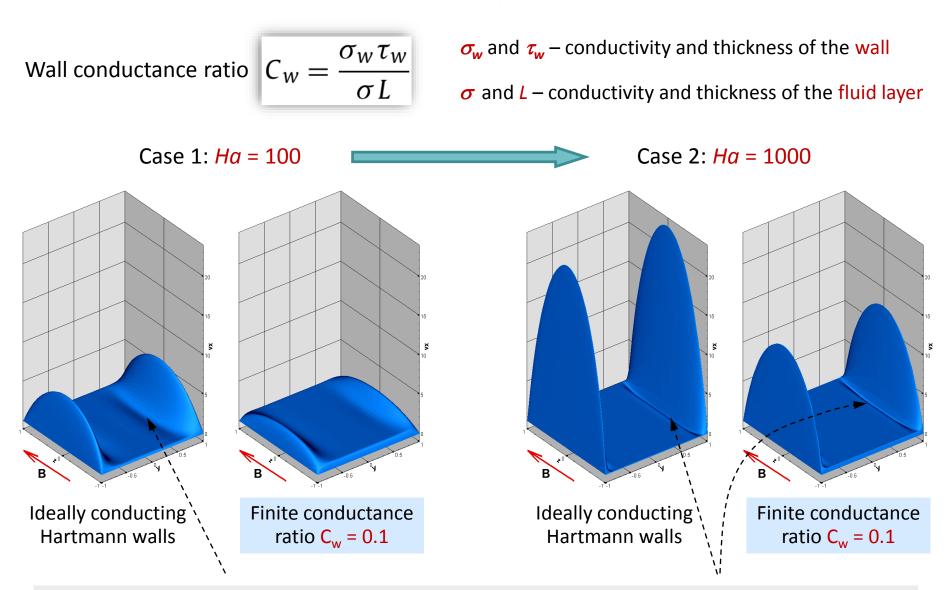
Archetypal system: MHD flow in square duct



• Boundary condition for electric potential, idealized cases

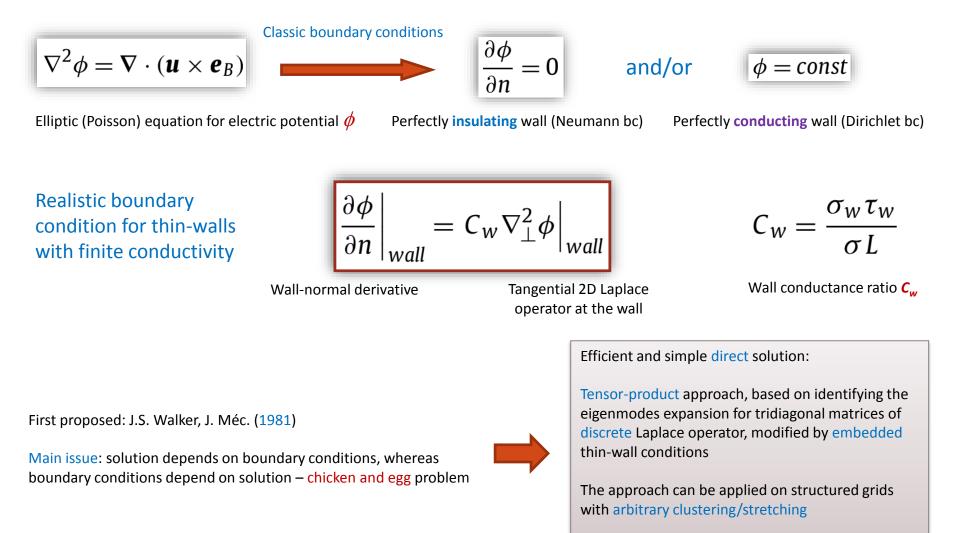
Side wall	$\partial \varphi / \partial n = 0$	$\partial \varphi / \partial n = 0$
Hartmann wall	$\partial \varphi / \partial n = 0$	$\varphi = const$
	perfectly insulating	perfectly conducting

Finite-wall conductivity (realistic scenario)



Importance: at high *Ha* the basic velocity profile develops inflection points in the region of side-jets – the flow, therefore, may (and do!) become unstable even at very low-*Re*

Tensor-product Elliptic solver for finite-wall conductivity



What is tensor-product approach for elliptic equations?

Tensor-product approach can be seen as the most generalized form of Fourier expansion, i.e. expansion into eigenmodes

$$\frac{\delta^2 \phi}{\delta x^2} + \frac{\delta^2 \phi}{\delta y^2} + \frac{\delta^2 \phi}{\delta z^2} = r$$



Each of these operators $\frac{\delta^2 \varphi}{\delta x^2}$, $\frac{\delta^2 \varphi}{\delta y^2}$, $\frac{\delta^2 \varphi}{\delta z^2}$ can be represented as tridiagonal matrix for the 3-point stencils of 2nd derivative

$$\frac{\delta^2}{\delta x^2} = \frac{1}{h^2} \begin{bmatrix} -2 & \cdots & \cdots \\ 1 & -2 & 1 \\ \cdots & \cdots & -2 \end{bmatrix}$$

Tridiag. matrix T_x for x-uniform grid

(1) Core idea – find eigenvalues λ , eigenvectors and inverse eigenvectors for tridiagonal matrices T_x and T_y corresponding to horizontal operators in x- and y-directions:

 $T_x \rightarrow A$ (direct), λ_x and A^{-1} (inverse) and $T_y \rightarrow B$ (direct), λ_y and B^{-1} (inverse)

(2) 3D elliptic problem can be converted to the eigenmodes space as series of 1D problems amenable to fast Thomas tridiagonal solver in the remaining z-direction

$$\left(\lambda_{x,i}+\lambda_{y,j}+\frac{\delta^2}{\delta z^2}\right)\hat{\phi}_{i,j}(z)=\hat{r}_{i,j}(z),\ i=1,\ldots,N_x,\ j=1,\ldots,N_y$$

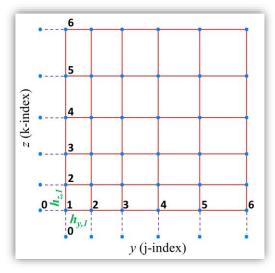
For uniform grids this approach retains full compatibility with fast transforms (FFT, cosFT, sinFT)!!!

The eigenvectors *A*, *B* converge to

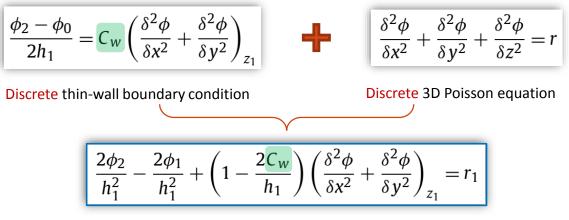
Fourier modes for periodic conditions Cosine modes for Neumann conditions Sine modes for Dirichlet conditions

The eigenvalues λ_x and λ_y become $-k_x^2$ and $-k_y^2$, i.e. negatives of the square of the wave-numbers

Embedding thin-wall b.c. into eigenmode expansion



Example of rectangular grid in (y,z)



This form of thin-wall condition can be implemented with eigenmodes expansion

Two methods are possible, both amount to modification of the tridiagonal matrices

Method 1. Solving problem in transformed (eigenmodes) space, corner elements of the tridiag. matrix for Thomas solver are modified

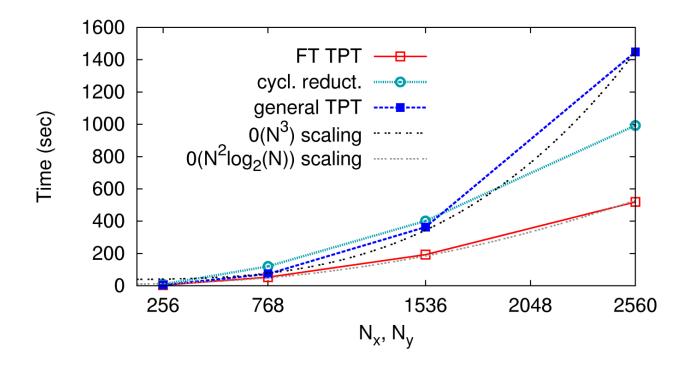
$$\frac{2\hat{\phi}_2}{h_1^2} - \frac{2\hat{\phi}_1}{h_1^2} + \left(1 - \frac{2C_w}{h_1}\right) \left(\lambda_{x,i} + \lambda_{y,j}\right) \hat{\phi}_1 = \hat{r}_1 \quad \Longrightarrow \quad [a_1, b_1, c_1] = \left[\times, -\frac{2}{h_1^2} + \left(1 - \frac{2C_w}{h_1}\right) \left(\lambda_{x,i} + \lambda_{y,j}\right), \frac{2}{h_1^2}\right]$$

Method 2. Modify tridiag. matrix in real space, then use this matrix to identify modified eigenmodes, then transform and solve

$$\frac{2\phi_2}{h_1^2}\frac{h_1}{h_1 - 2C_w} - \frac{2\phi_1}{h_1^2}\frac{h_1}{h_1 - 2C_w} + \left(\frac{\delta^2\phi}{\delta x^2} + \frac{\delta^2\phi}{\delta y^2}\right)_{z_1} = r_1\frac{h_1}{h_1 - 2C_w} \implies [a_1, b_1, c_1] = \left[\times, -\frac{2}{h_1^2}\frac{h_1}{h_1 - 2C_w}, \frac{2}{h_1^2}\frac{h_1}{h_1 - 2C_w}\right]$$

Benchmarks of tensor-product vs. other methods (performed on SuperMUC-NG)

Rayleigh-Bénard convection in a rectangular box computed on grids with various $N_x = N_y = N$ and $N_z = 256$ Clock time required to compute 100 time steps on 256 cores is shown as a function of N Matrix multiplications in general TPT are performed with MKL multi-thread routines



General TPT =Tensor-Product-Thomas solver, i.e. MatMul in x,y and Thomas method in zCycl. Reduct. =Cosine-FT in x and 2D Cyclic reduction method (Fishpack) in y,zFT TPT =Fast Transform Tensor-Product-Thomas, i.e. Cosine-FT in x,y and Thomas method in z

Only TPT allows for arbitrary grid-clustering in x,y,z

#2: Tensor-product elliptic solver, list of pros and cons

(in part inspired by car reviews 😊 from Mat Watson/carwow youtube)

Prerequisite – elliptic problem should be separable, thus decomposition into eigenmodes can be applied (basically that implies constant coefficients and structured grids)

Good points

- Direct solver no iterations, no convergence issues
- Arbitrary grid-clustering in all 3 directions
- MKL matmul routines scale up almost linearly vs. number of threads
- Can be extended to curvilinear grid if separable (e.g. cylinder coords.)
- No speed-penalty vs. classic Neumann and Dirichet conditions
- Each wall can be assigned its own conductance ratio C_w
- Can be applied for finite thermal conductivity, even for unsteady form

Annoying points

- Transform matrices grow in size as n^2 potential issue at hi-res
- Matrix multiplication scales as O(n³), vs. O(n²log₂(n)) for FFT/CosFT/SinFT

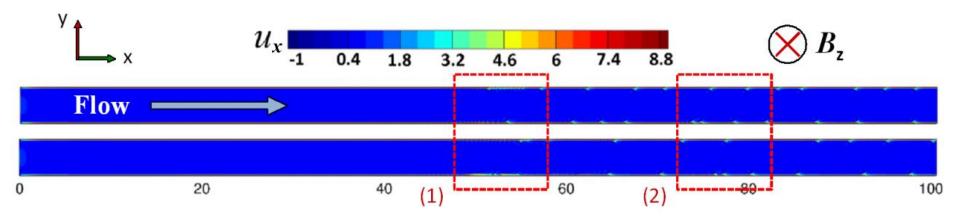
Five good points about the car Five annoving points about the car

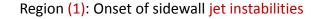


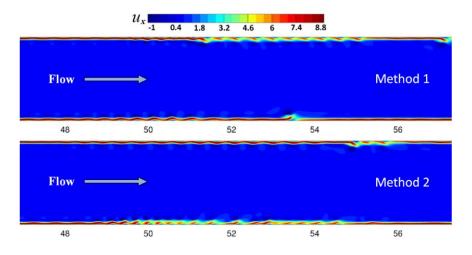
Copyright carwow.co.uk and Mat Watson

#3: Evolution of MHD flows in ducts and rectangular boxes (examples)

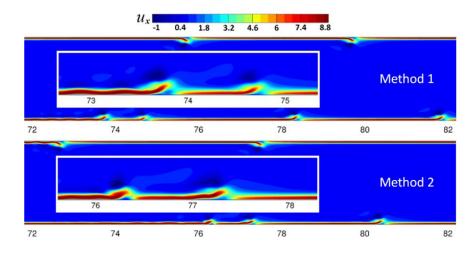
MHD flows in ducts with conducting walls – typical configuration for liquid-metal fusion blankets Spatial evolution of <u>Hunt's</u> flow at Re = 2000, Ha = 2000 and $C_w = 0.03$







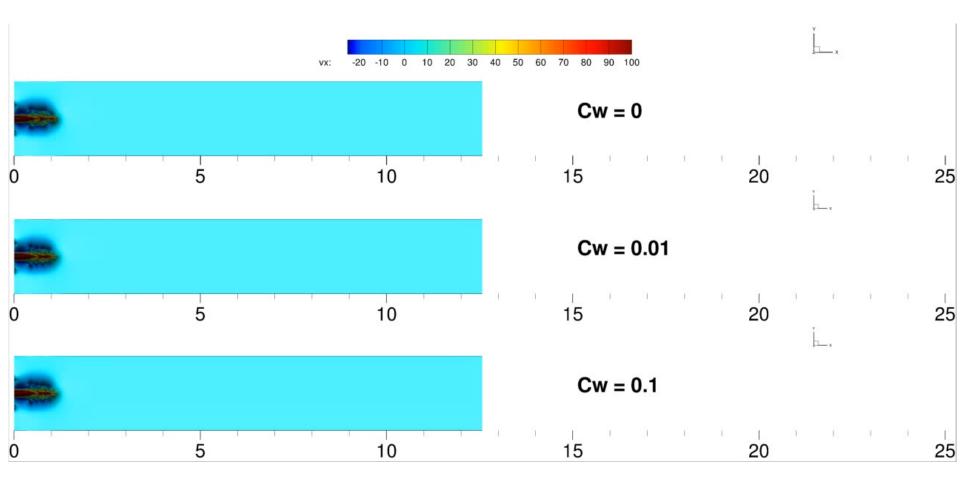
Region (2): Fully developed jet detachments



• Krasnov et.al. J. Comp. Phys. (2023)

#3: Evolution of MHD flows in ducts and rectangular boxes (examples)

Effect of wall conductivity on instability of a submerged <u>round jet</u>, entering square duct Uniform vertical magnetic field B_2 at <u>Ha</u>=500, flow at <u>Re</u>=1000

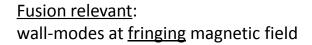


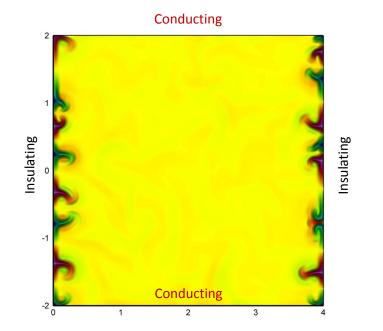
• Belyaev et.al., J. Fluid. Mech. (2023)

#3: Evolution of MHD flows in ducts and rectangular boxes (examples)

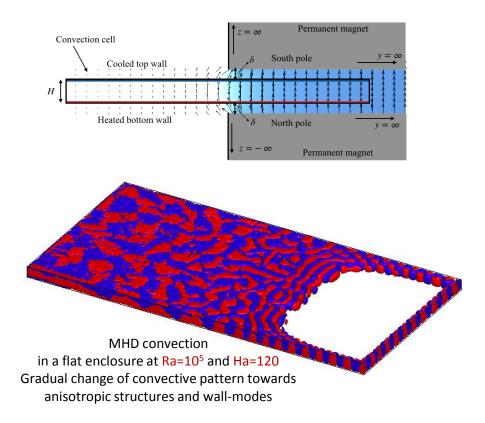
Magneto-thermal convection flows around and above the Chandrasekhar stability limit At these conditions convective motion is expected to be fully suppressed by strong magnetic field However, residual motion at the sidewalls – wall-modes – can exist far above even if the rest has "died"

<u>Fundamental</u>: effect of wall conductivity on the wall-modes at <u>uniform</u> magnetic field





MHD convection in a rectangular box at $Ra=10^7$ and Ha=1000Wall-modes are killed at conducting side-walls $C_w=0.1$



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Outlook & Possible next steps

- Convection at low *Pr*: exploring broader parameter space and longer statistics (time-evolution)
- Extension of the tensor-product approach to solve problems with finite thermal wall-conductivity → conjugate heat transport "made easy"
- It would be interesting to explore application of practical algorithms of fast matrix multiplication possibly utilizing the symmetry properties of the transform matrices to achieve scaling better than O(n³)

Thanks a lot for your attention!

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