

A Koopman operator approach for identifying
 $2D+1+1$
transport structures in HYCOM North Atlantic

Andrew Poje, CUNY-CSI

MURI 3D+1 Workshop - RSMAS

Miami, Nov 6-7, 2014

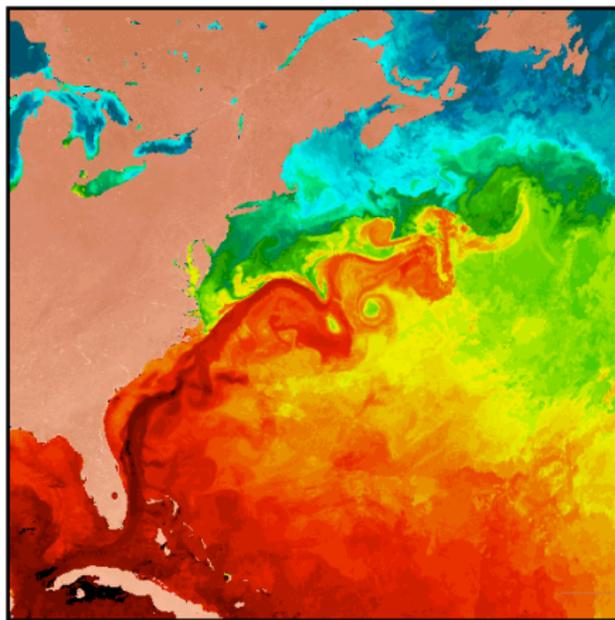
Supported by: Office of Naval Research Multiple University Research Initiative: Analysis of 3D+1 Ocean Flows

Collaborators:

- Rakulan Sundralingam, Lucas Garber, Alexandre Fabregat – CUNY Math/Phys
- Igor Mezić, Marko Budišić – UCSB
- Angelique Haza, Tamay Özgökmen – RSMAS
- MURI 3D+1 Team

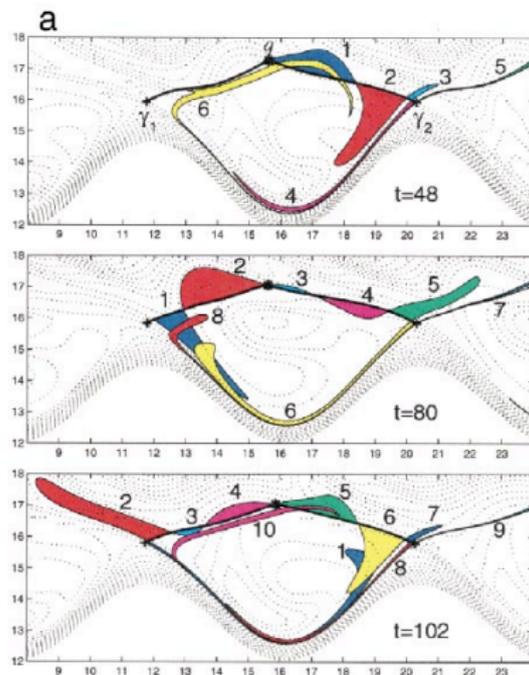
'Its deja-vu, all over again', L.P.B.

- LCS: Visualize time history of Lagrangian kinematics in Eulerian frame.
- Color/classify/delineate IC field based on common kinematics
- Define Lagrangian Structures
...
- use definition to calculate transport between structures.



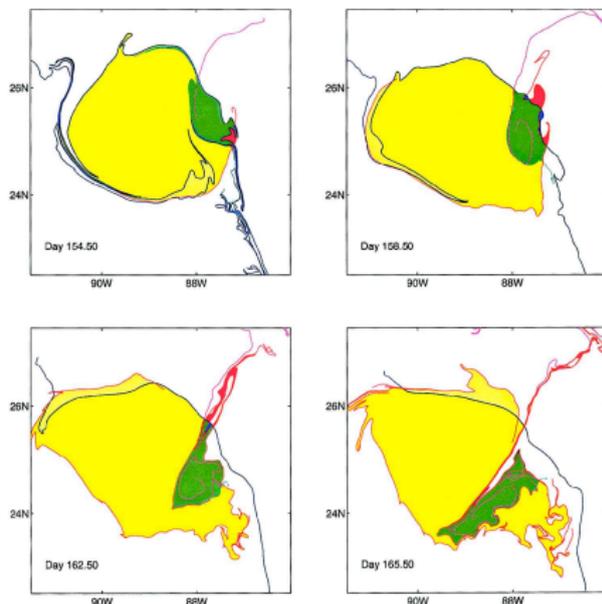
Overview: 1998 \Rightarrow 2014 (In two slides)

- Models are much better (or models have much higher resolution)
- 'Explosion of submesoscale energy' (Capet)
- Ageostrophic - $\nabla_h \mathbf{u} \sim 0$?
- Interest has shifted to smaller scales
- LCS: No longer chasing 'slow/fat', but 'fast/small'
- ... with vertical structure.



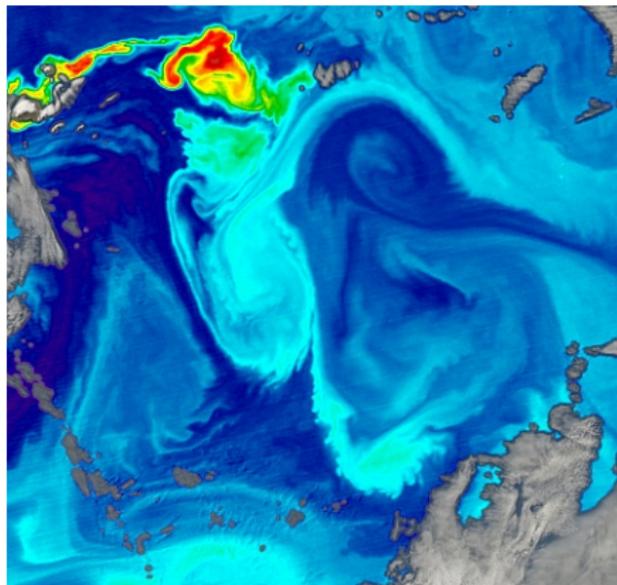
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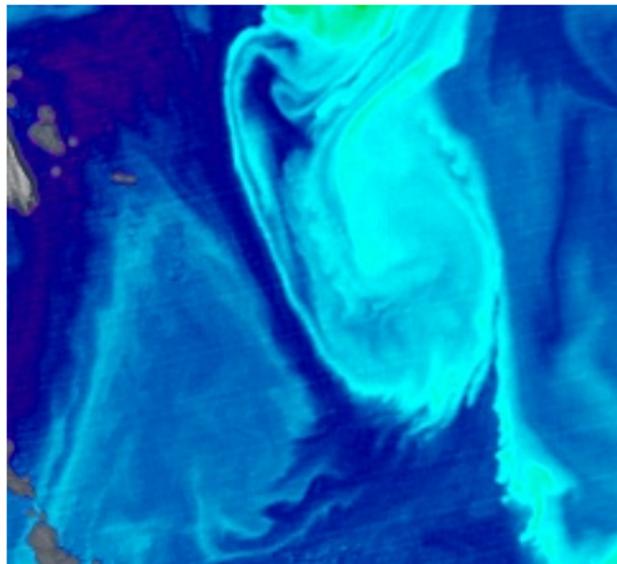
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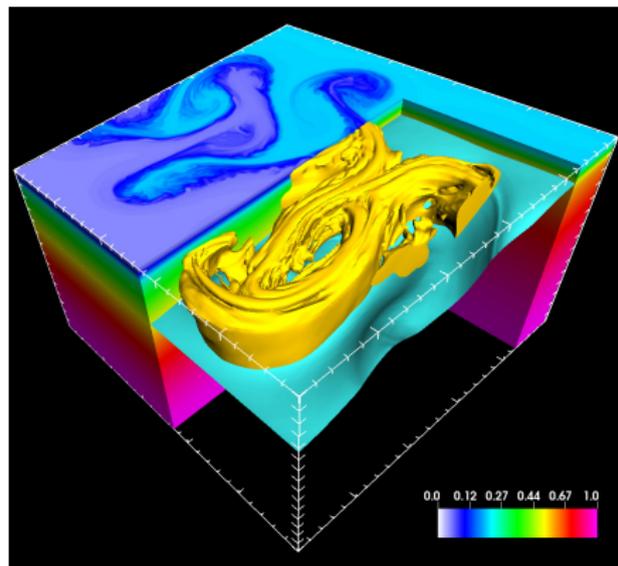
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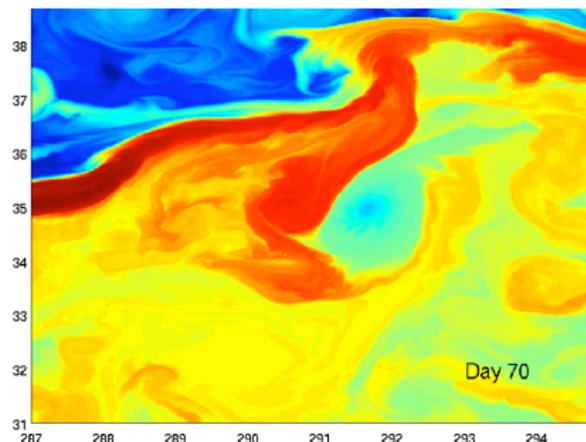
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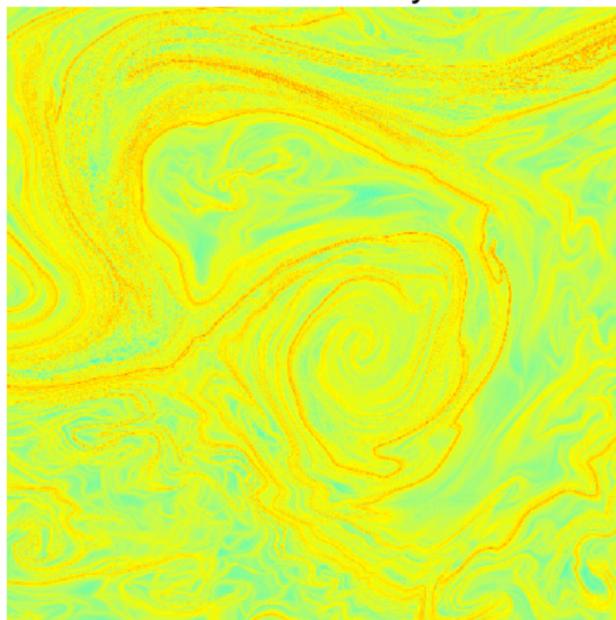
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unstable (hyperbolic) structures (boundaries)
- To: stable (elliptic) structures (cores)
- Realization: 'hyperbolic invariant manifolds are typically densely imbedded in structures' we want.



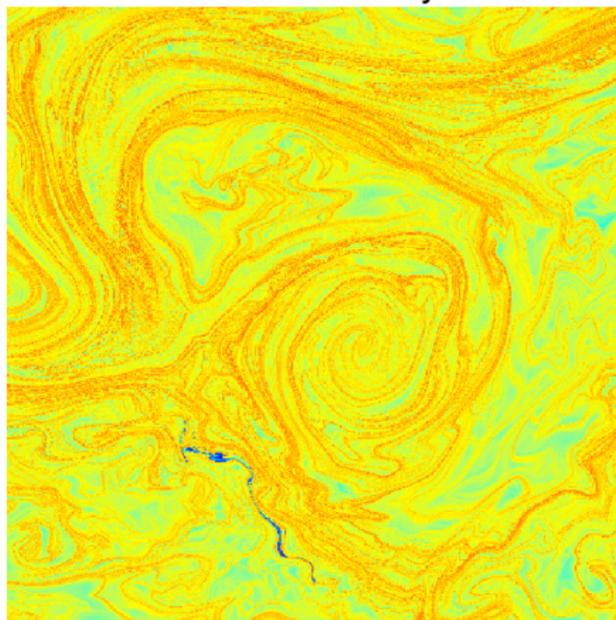
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DLE: $T = 8$ days



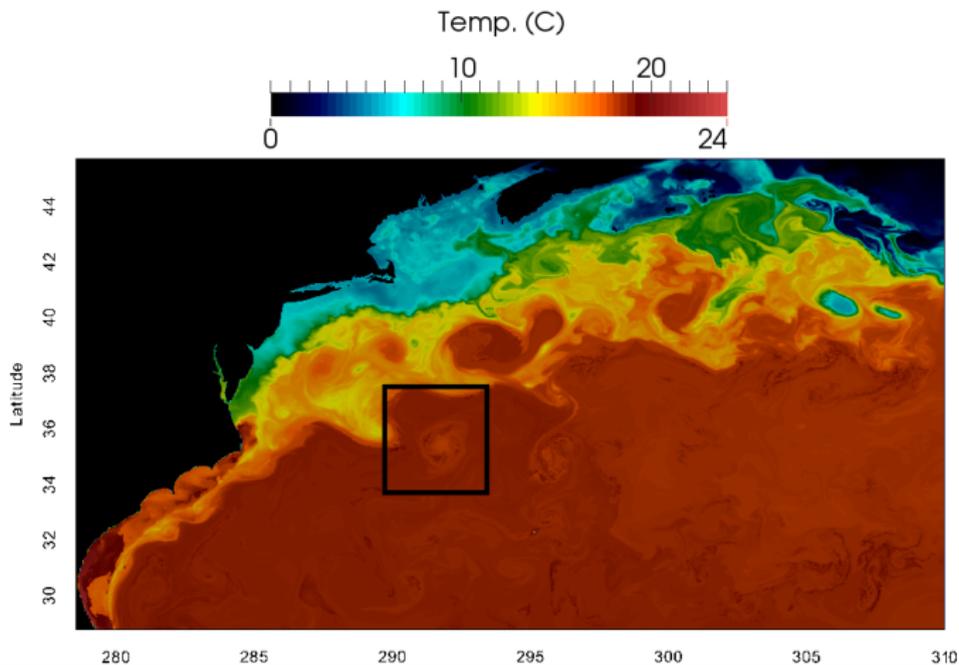
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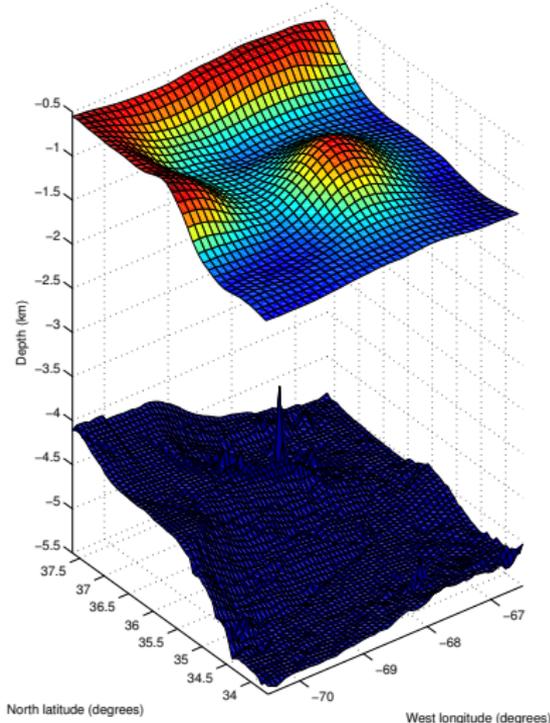
Goal: Visualize transport in 1/50° HYCOM North Atlantic

- Specifically: Dipole with strong, persistent Cold-Core Cyclone
- Eddy-jet, eddy-eddy interaction - Cross-Jet exchange
- 2D + 1 + 1: Advect on Isopycnals



Goal: Gulf-Stream Rings & HYCOM at 1/50 °

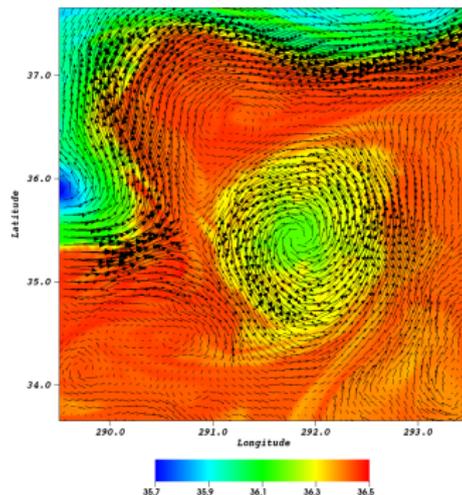
- HYCOM data-set: $\Delta x \sim 1.5$ km.
- 30 isopycnal layers (surface stacked).
- Gulf stream rings with realistic size, strength, frequency.
- Energy at sub-mesoscales (< 10 km) in model. (sub-mesoscale eddy permitting)
- Complex, small-scale vertical tracer structures.
- Database: 12hr fields (90,000 particles)



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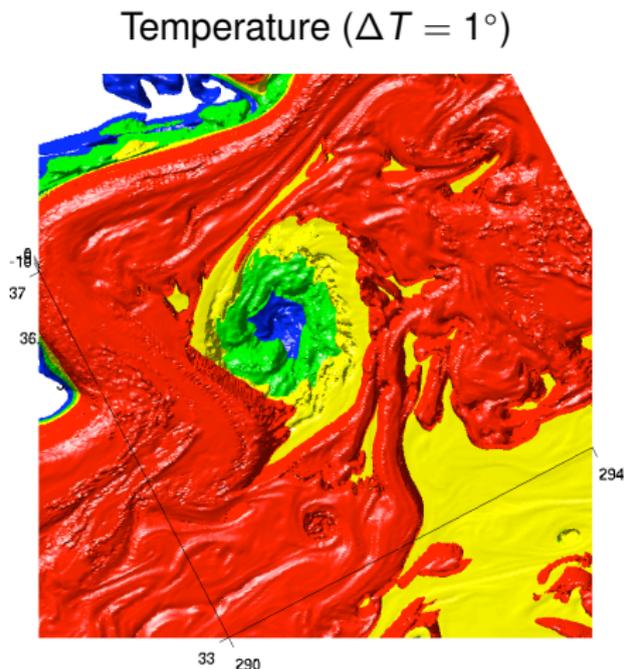
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Salinity



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- Koopman Formalism: (Budišić, Mohr & Mezić)

- ▶ Consider finite-time map (flow) $T : M \rightarrow M$
- ▶ Koopman Operator: $U_T : \mathcal{F} \rightarrow \mathcal{F}$

$$[U_T f] = f(T(p))$$

- ▶ Nonlinear ODES in $\mathbb{R}^n \rightarrow$ spectrum of *linear* operator U_T .
- ▶ Koopman modes from **Lagrangian averages** of specified basis set
- ...

$$f_k(x) = e^{2\pi i k \cdot x}$$

- ▶ Challenges in Ocean Context:

- ★ Open flow: M essentially non-compact $T : M \rightarrow M'$, $M \subset M'$.
- ★ Finite time: No well defined averaging time: $\mu(M') = g(t)$
- ★ Open flow and/or Finite time \rightarrow well defined partition? No strictly *invariant sets*.

- Chose set of *observables*:

$$f_k(\mathbf{x})$$

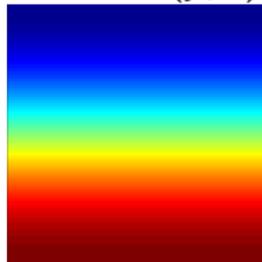
- Examine:

$$\overline{f_k(\mathbf{x}_0)}_\tau = \frac{1}{\tau} \int_0^\tau f_k(T(\mathbf{x}_0)) dt$$

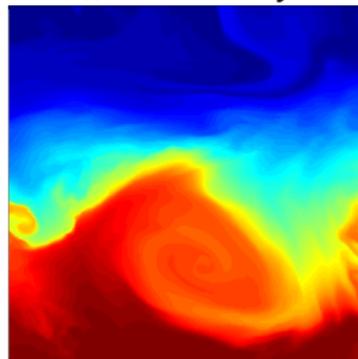
- For finite-time, non-time-periodic flows, chose

- ▶ Time-scale via τ .
- ▶ (Nominal) length-scale via $f_k(\mathbf{x})$.

$$f_1 = -\sin(y/2)$$



$$\overline{f_1}, \tau = 04 \text{ days}$$



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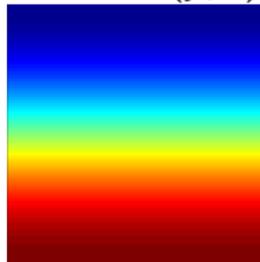
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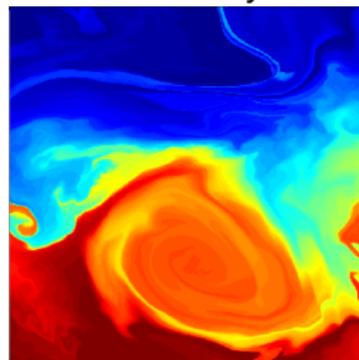
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$\tau = 08$ days



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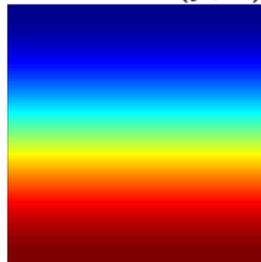
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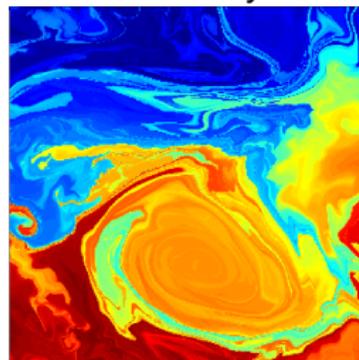
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$$\tau = 16 \text{ days}$$



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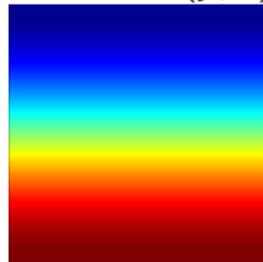
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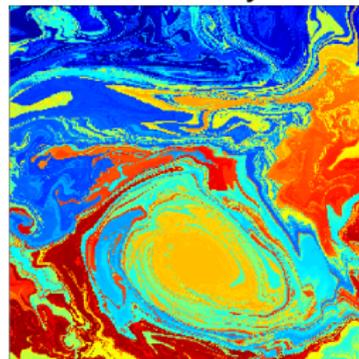
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$$\tau = 32 \text{ days}$$

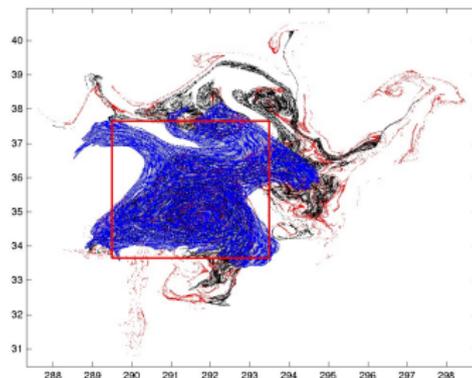


Depth dependent *Observables*

- Issue: Disparate average \mathbf{v} with depth.
- Disparate average trajectory length.
- Fix: Depth dependent averaging times.

$$\tau = \tau(\Delta\rho)$$

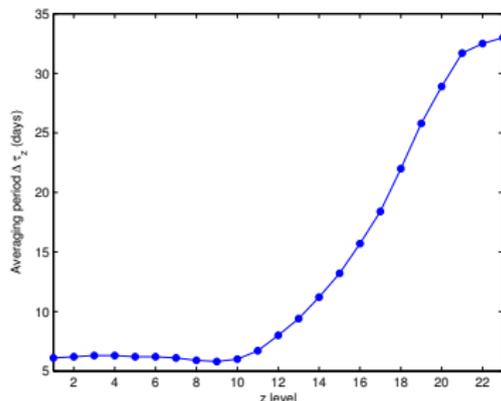
T = 7 days: 3 Layers



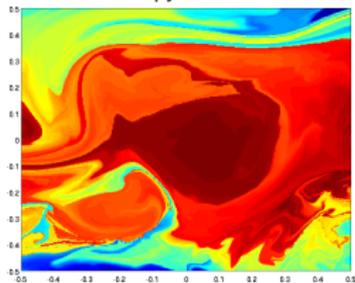
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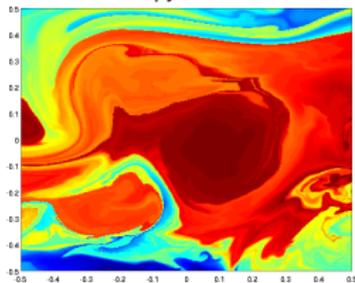
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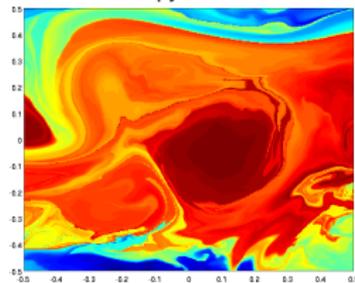
Isopycnal 1



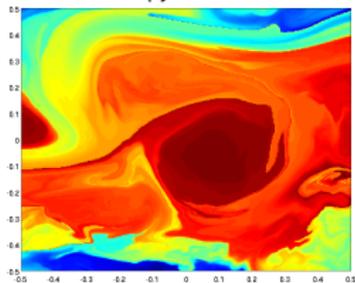
Isopycnal 4



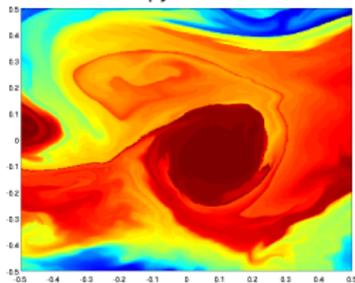
Isopycnal 7



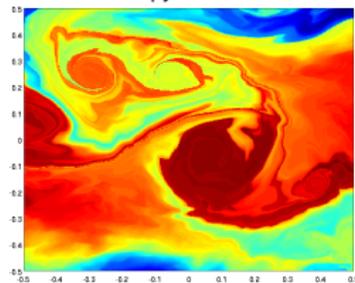
Isopycnal 10



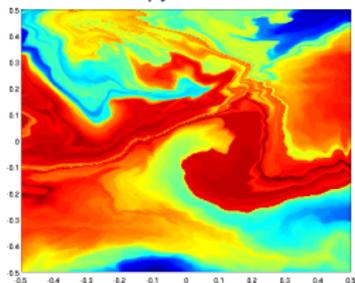
Isopycnal 13



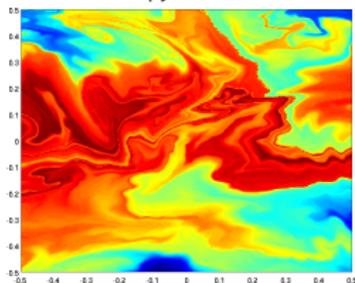
Isopycnal 16



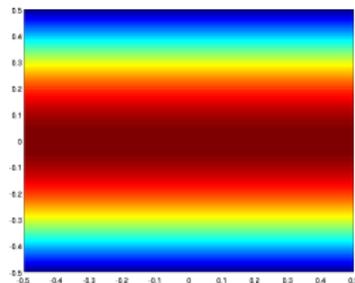
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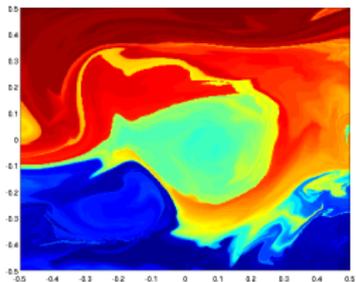
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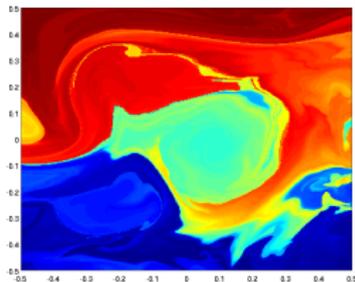
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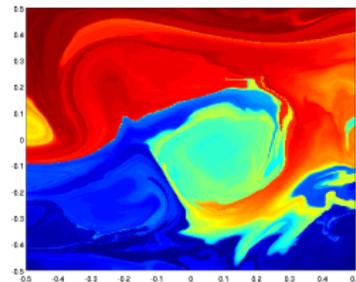
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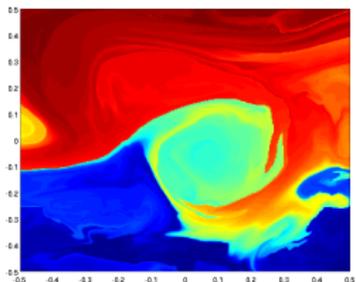
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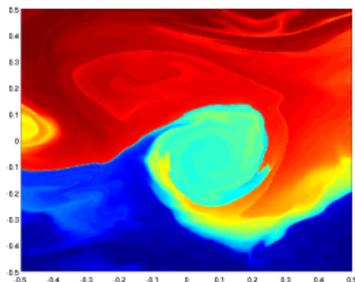
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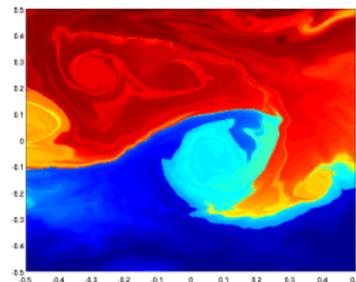
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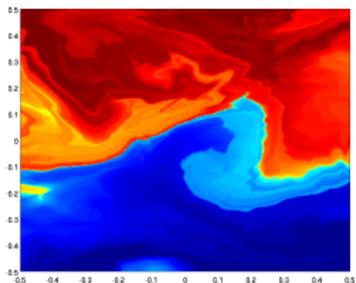
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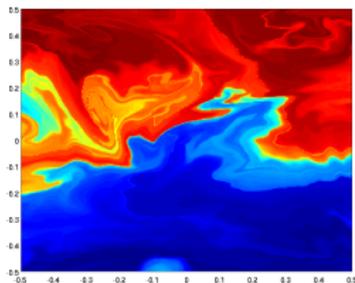
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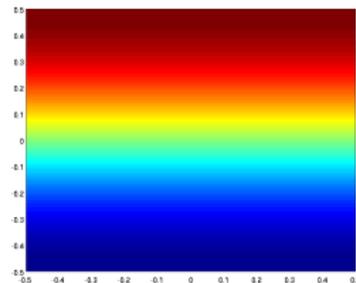
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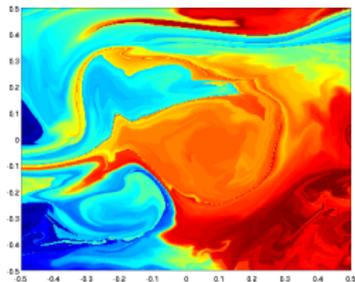
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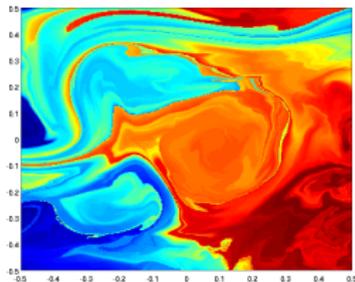
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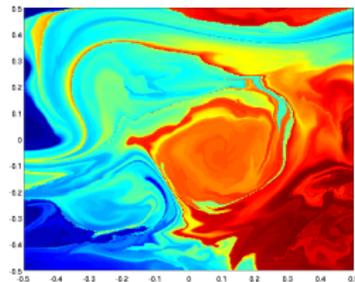
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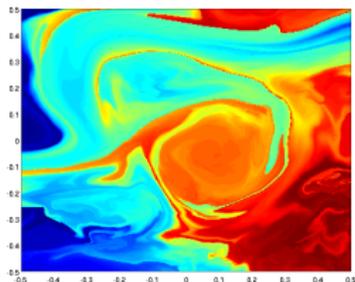
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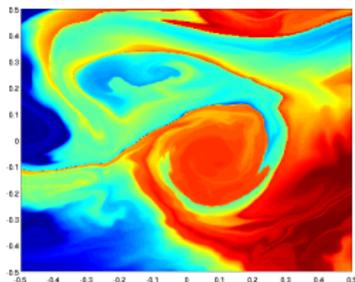
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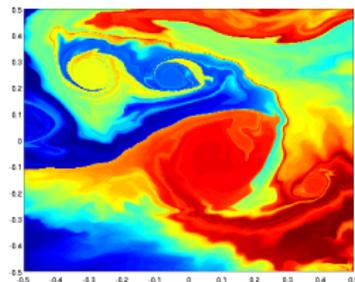
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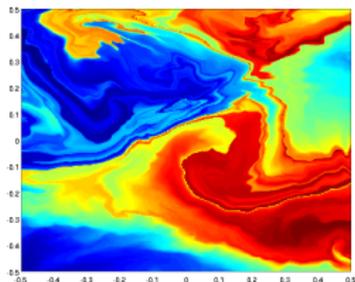
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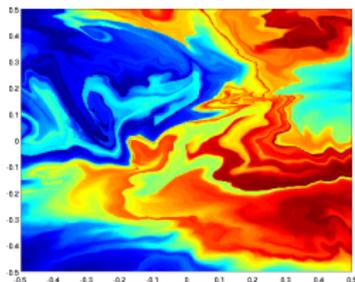
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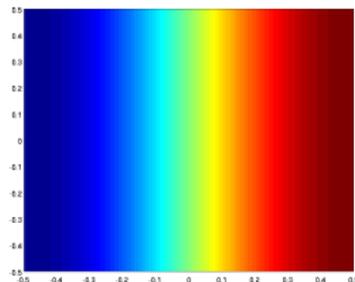
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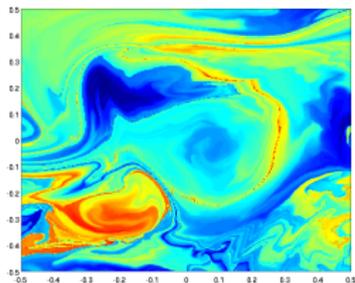
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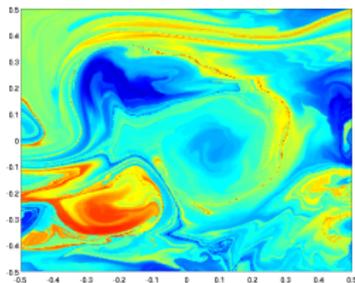
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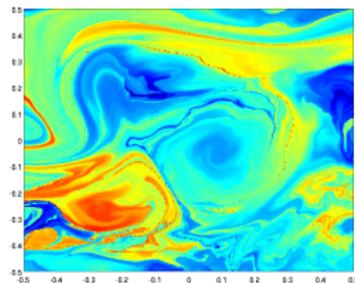
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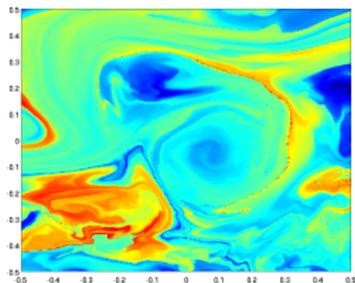
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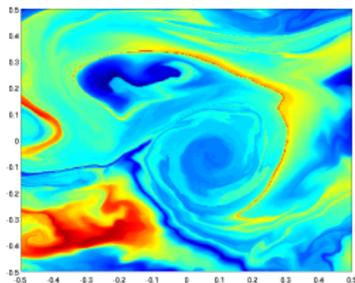
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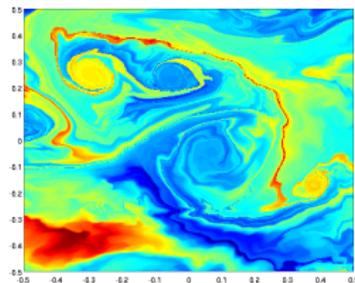
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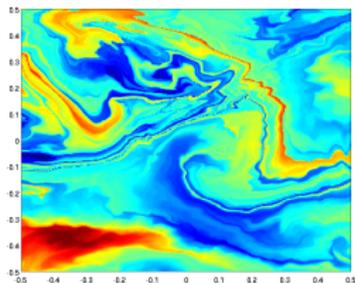
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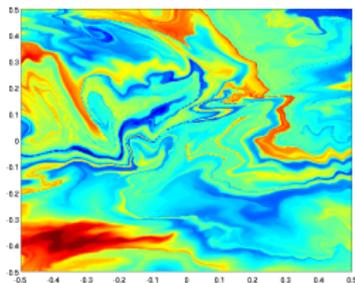
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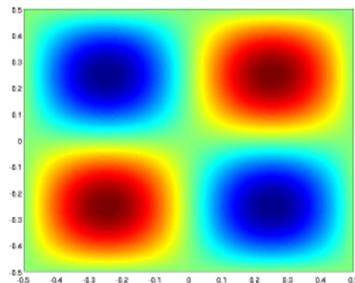
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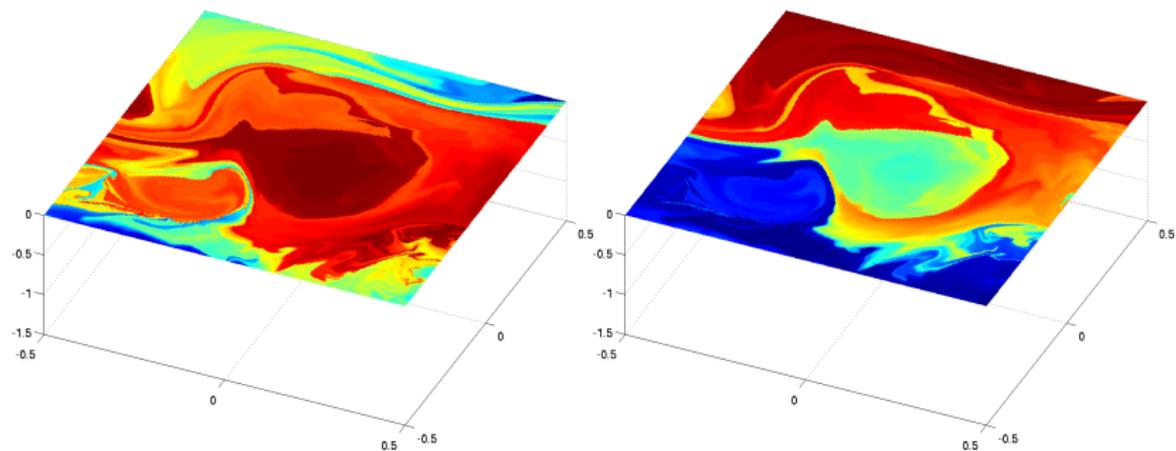
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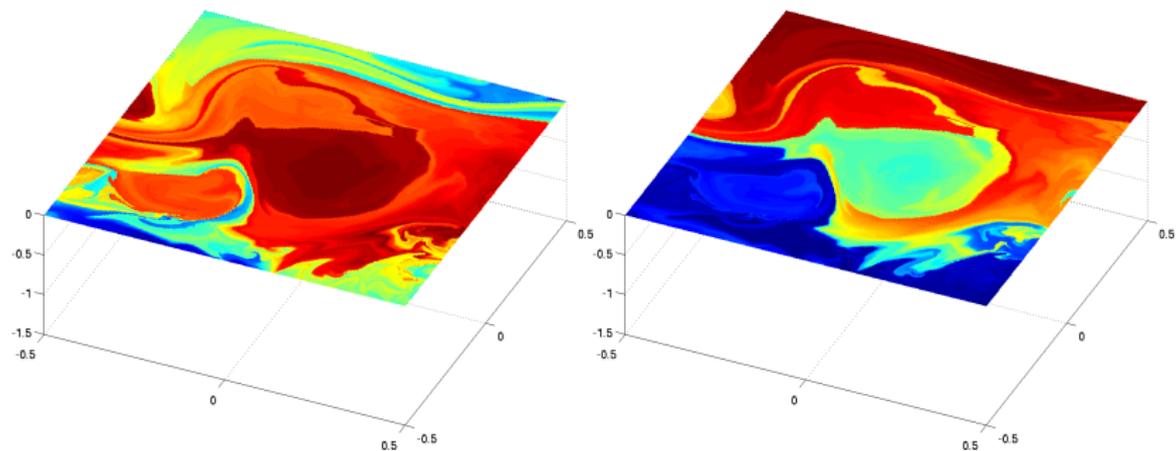
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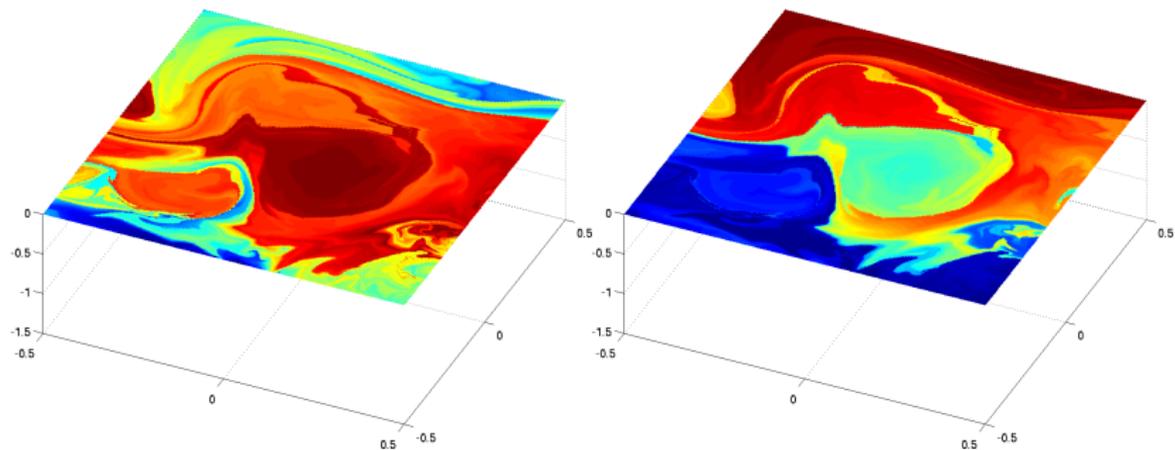
Depth dependence: Observable $k_x = 0$, $k_y = 1/2$



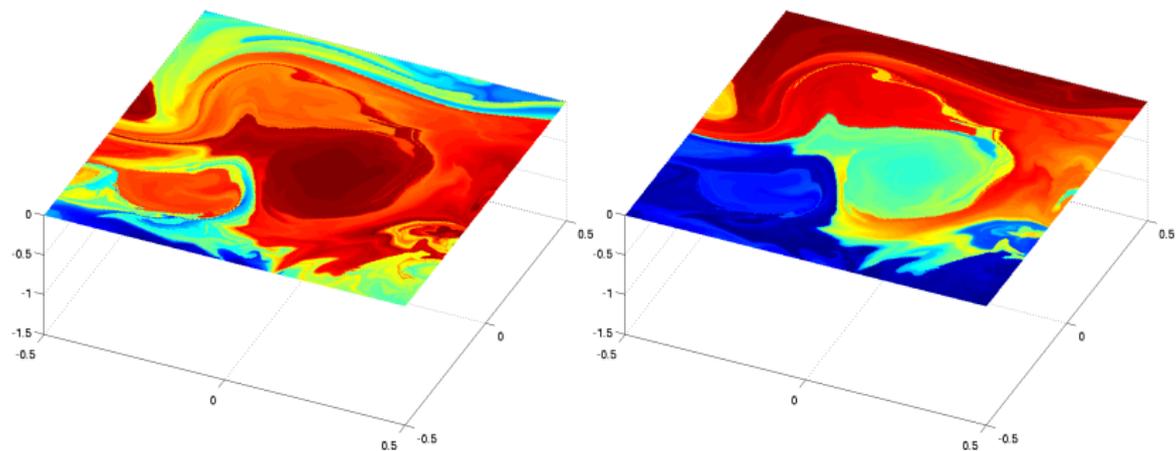
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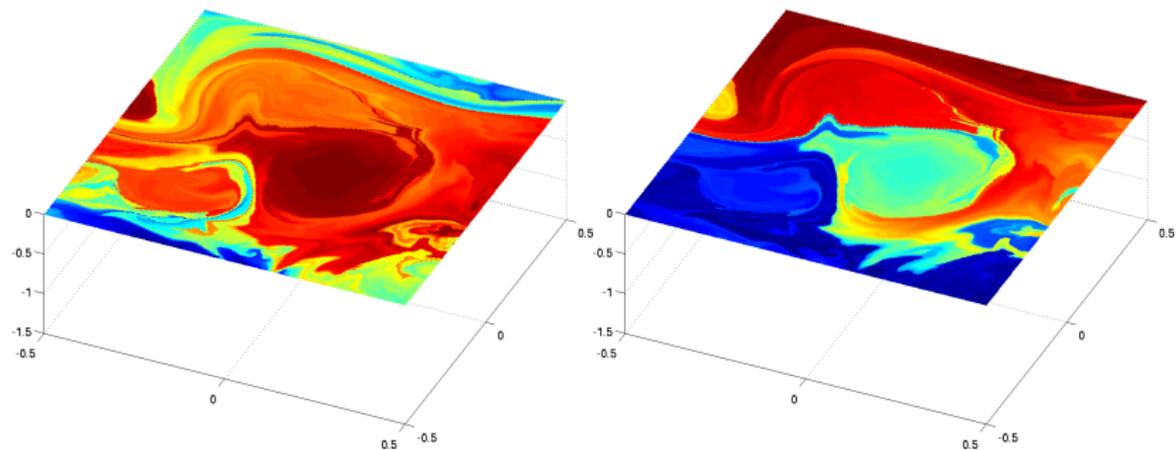
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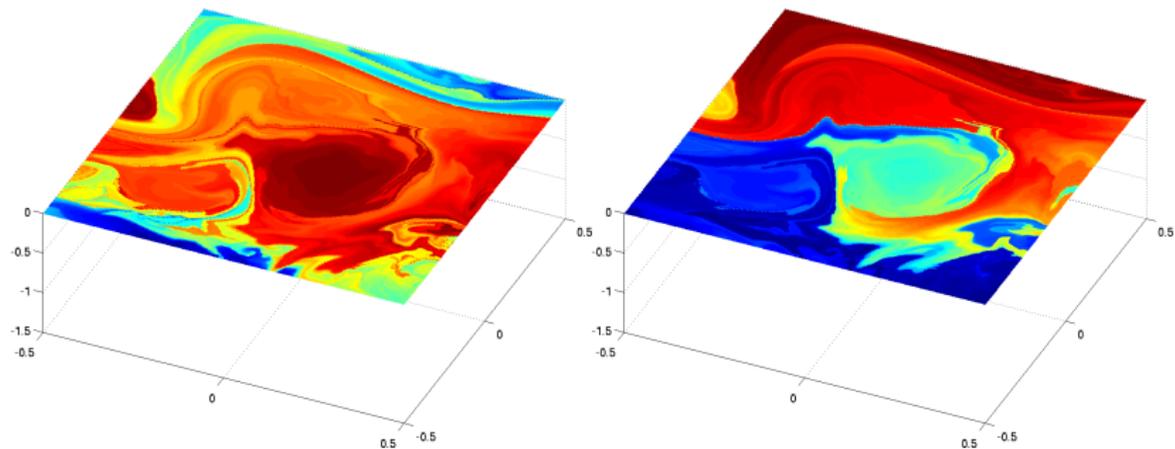
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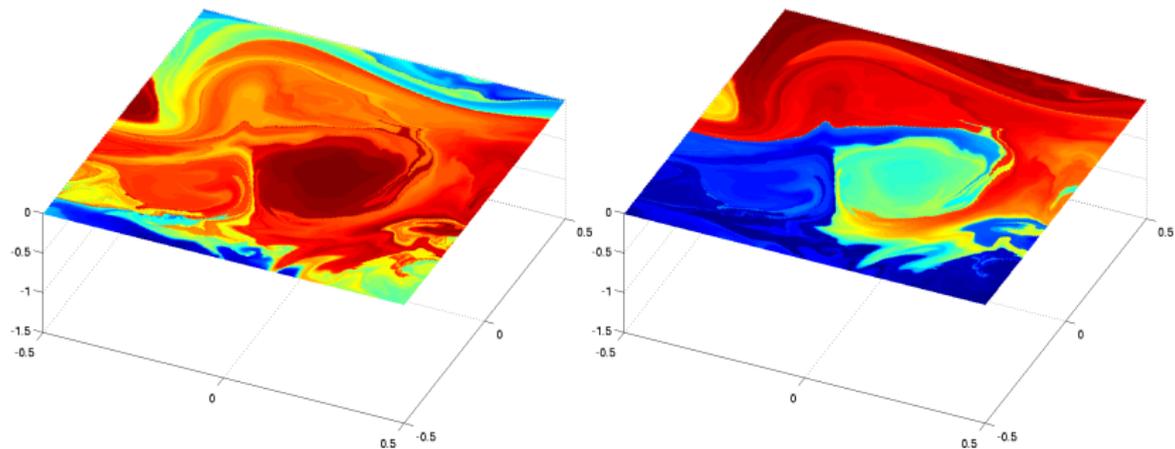
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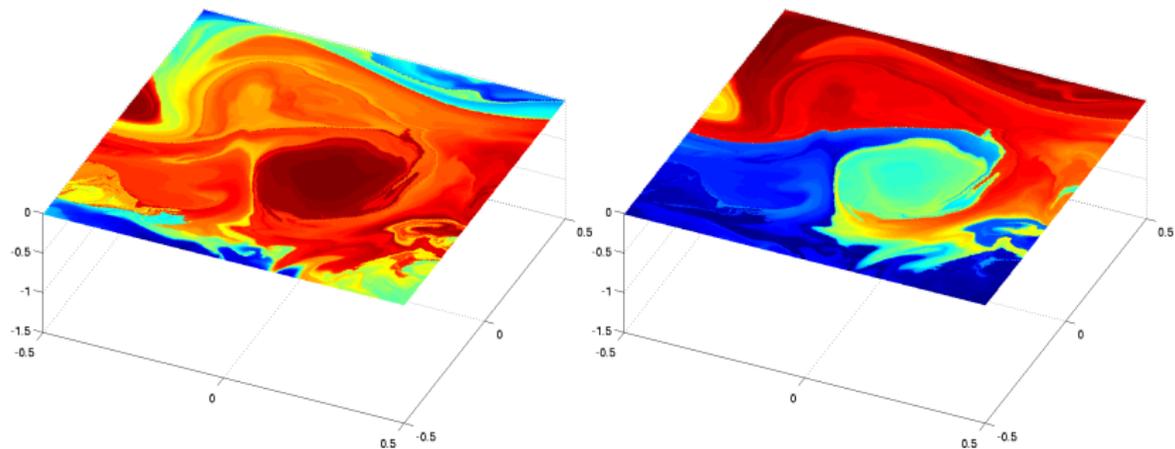
Depth dependence: Observable $k_x = 0$, $k_y = 1/2$



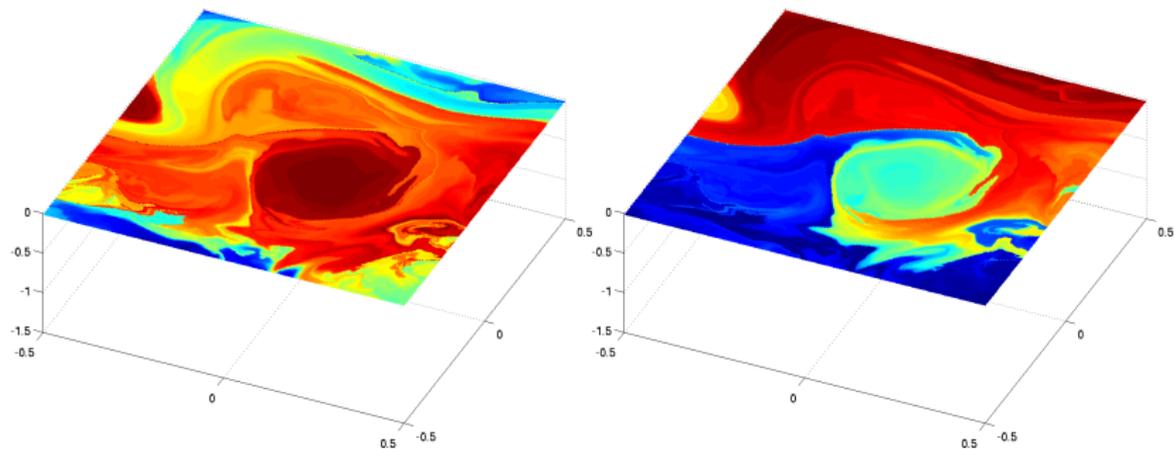
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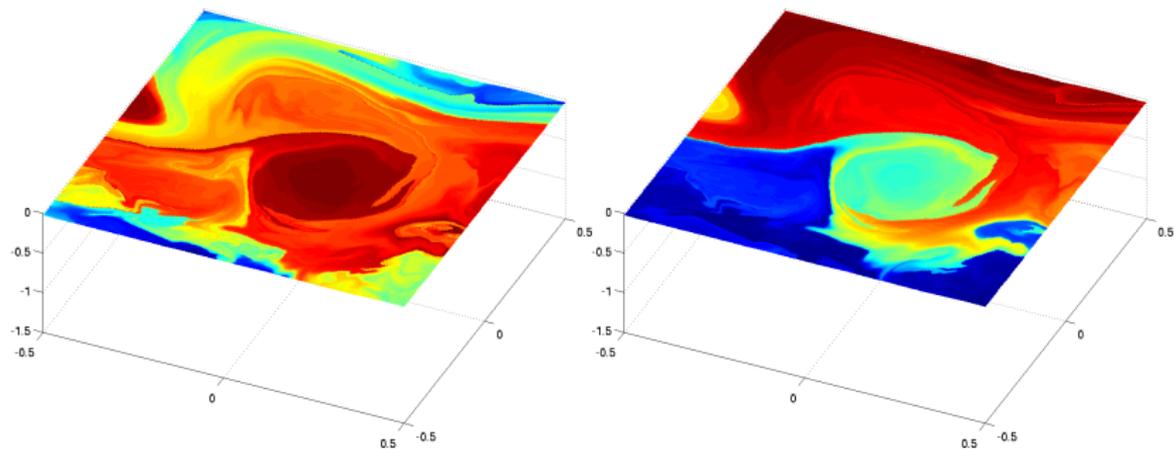
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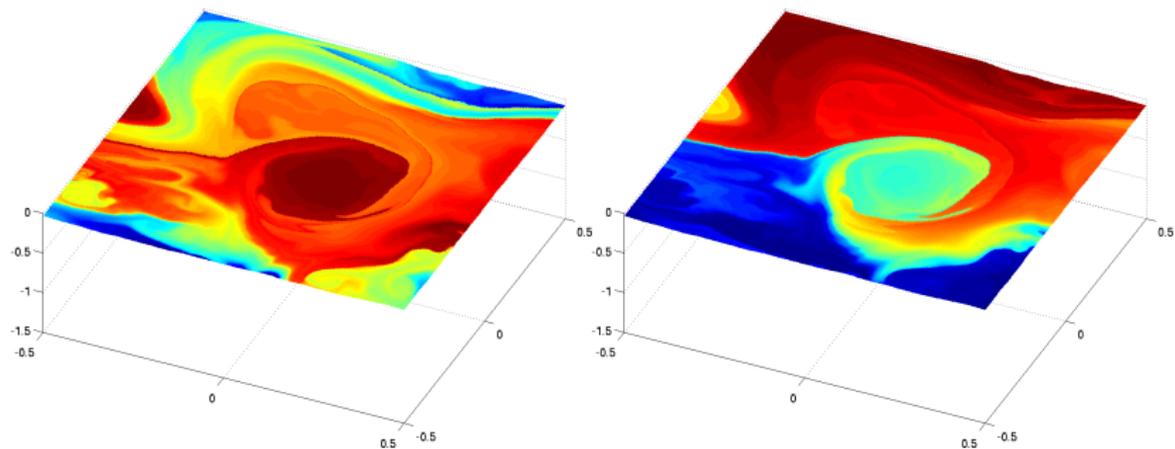
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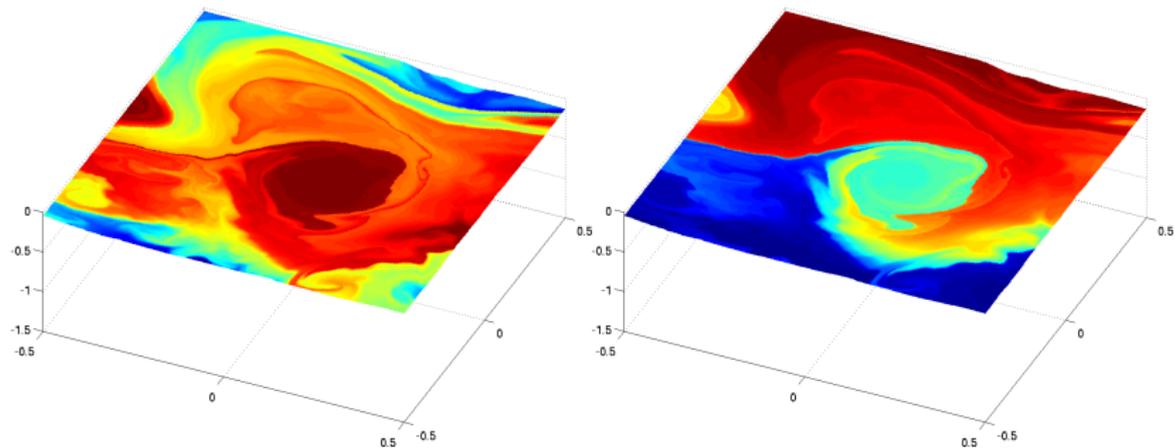
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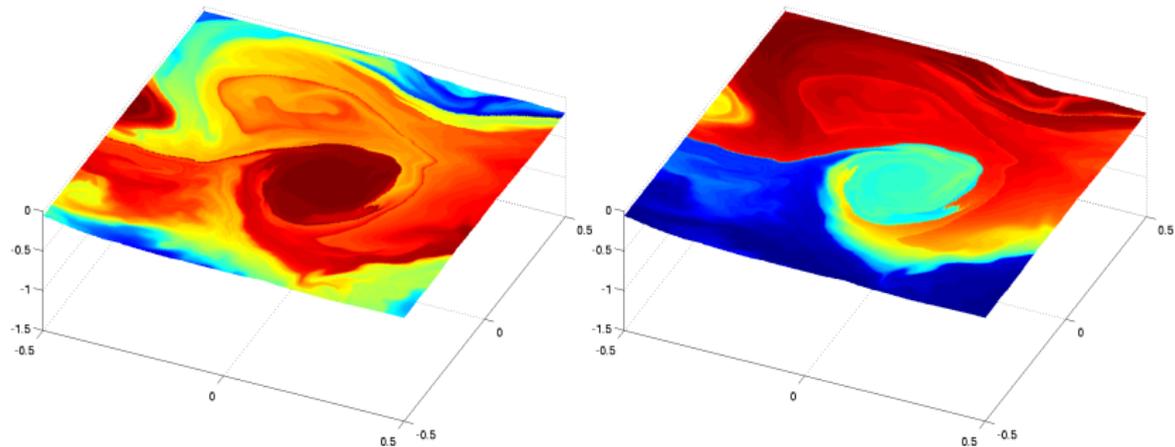
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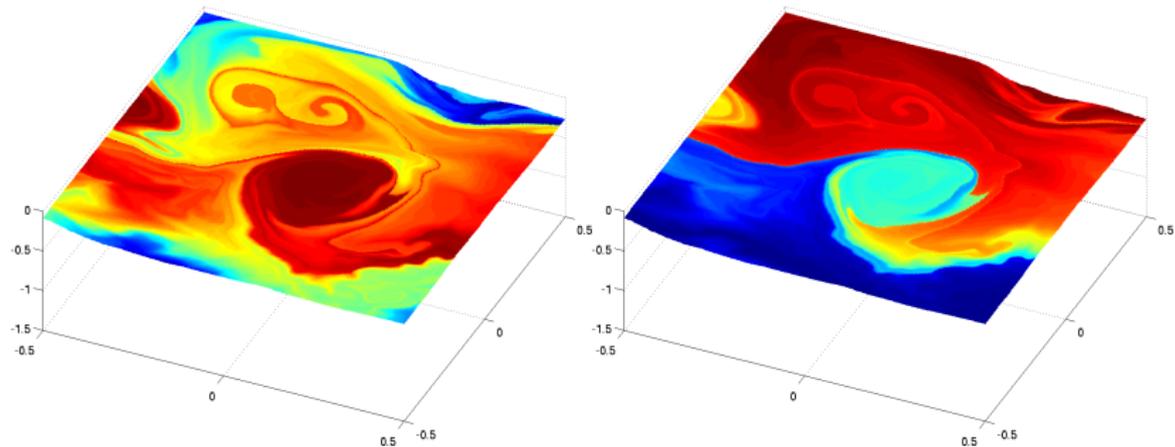
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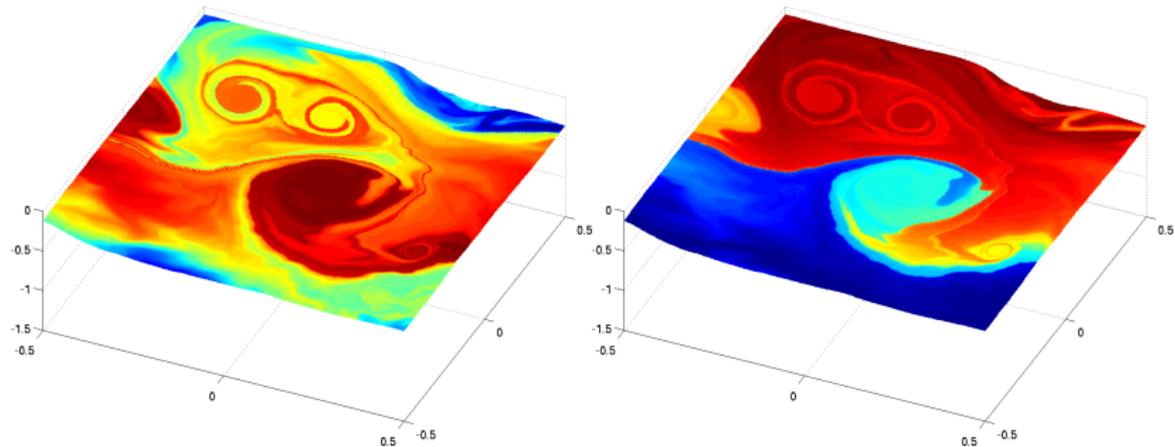
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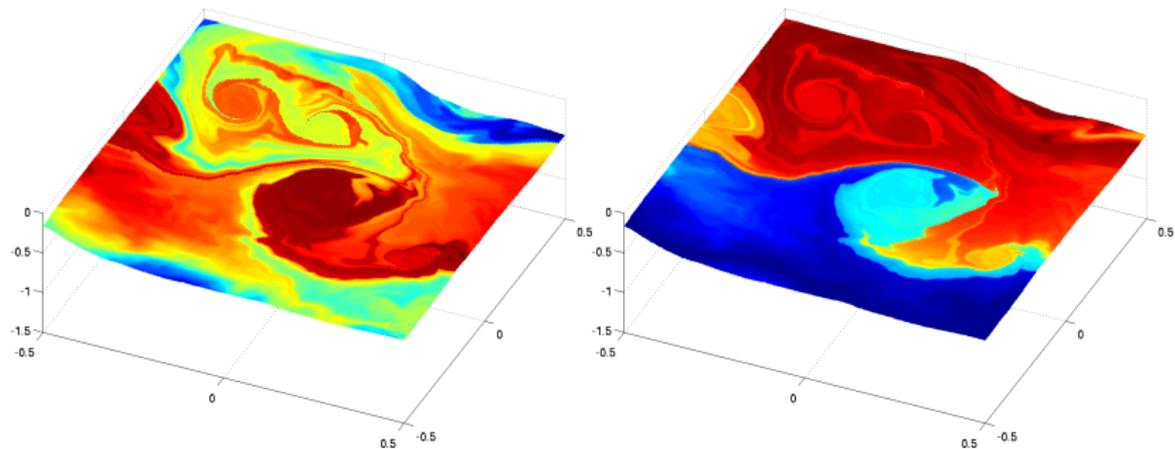
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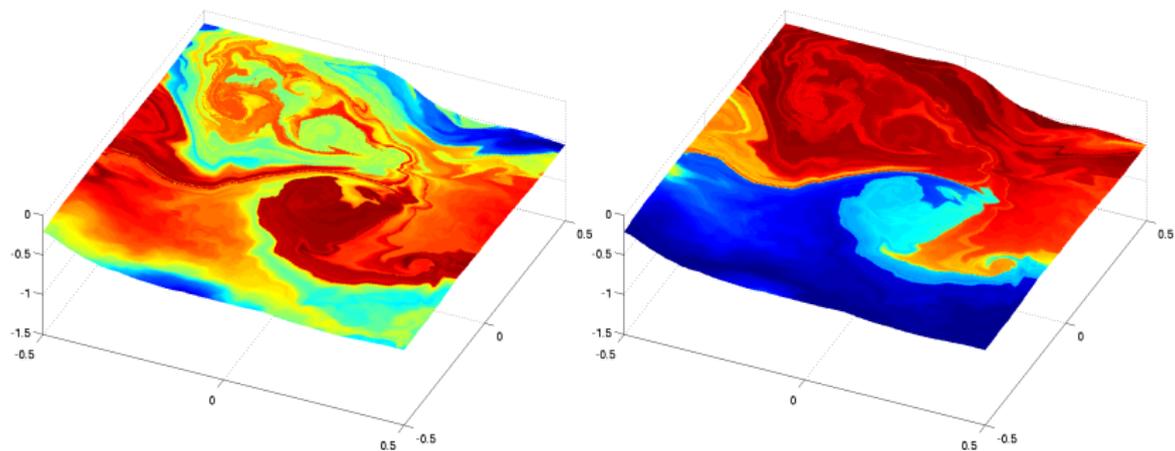
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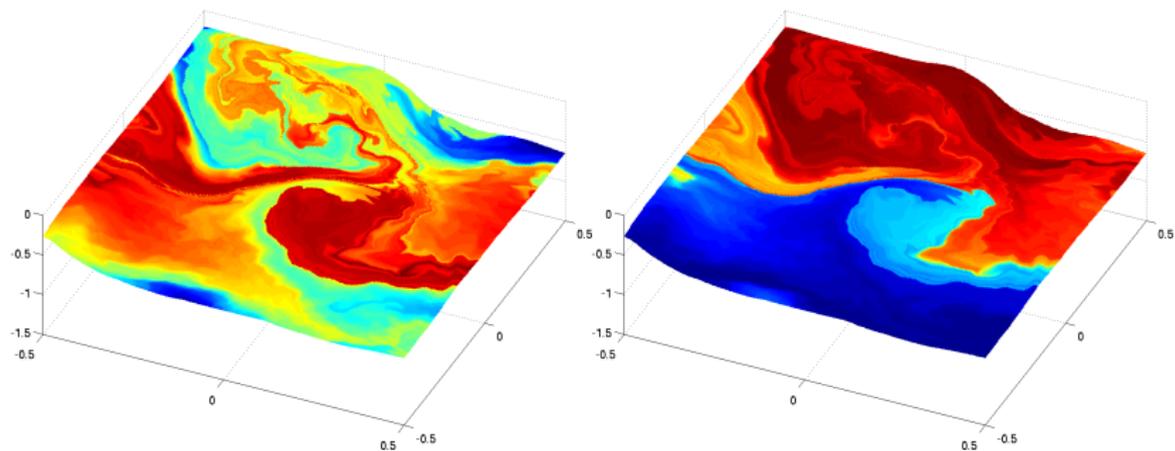
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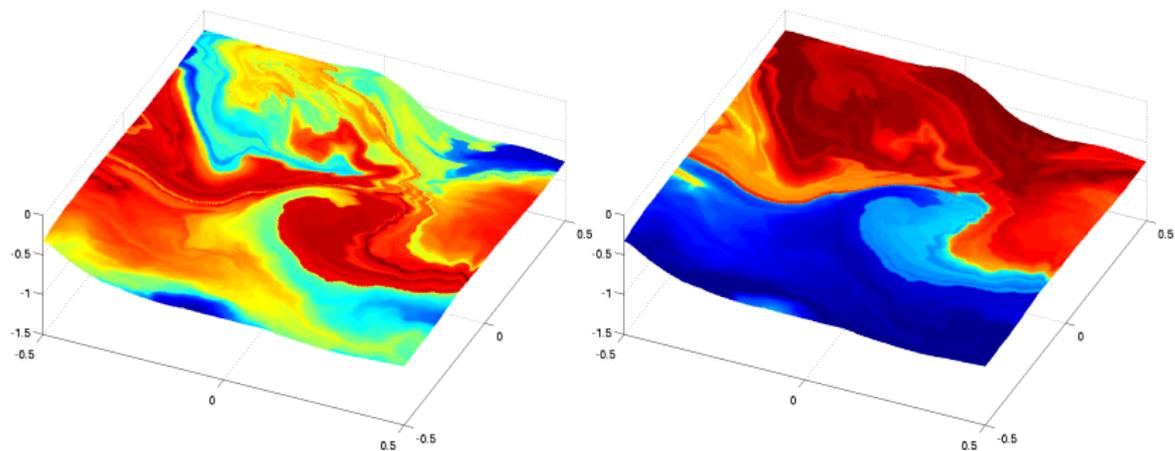
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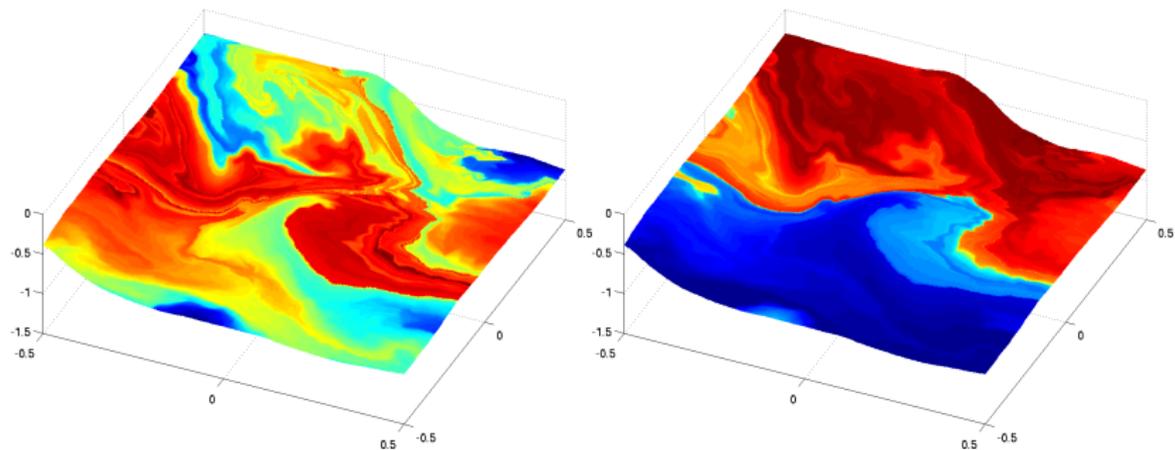
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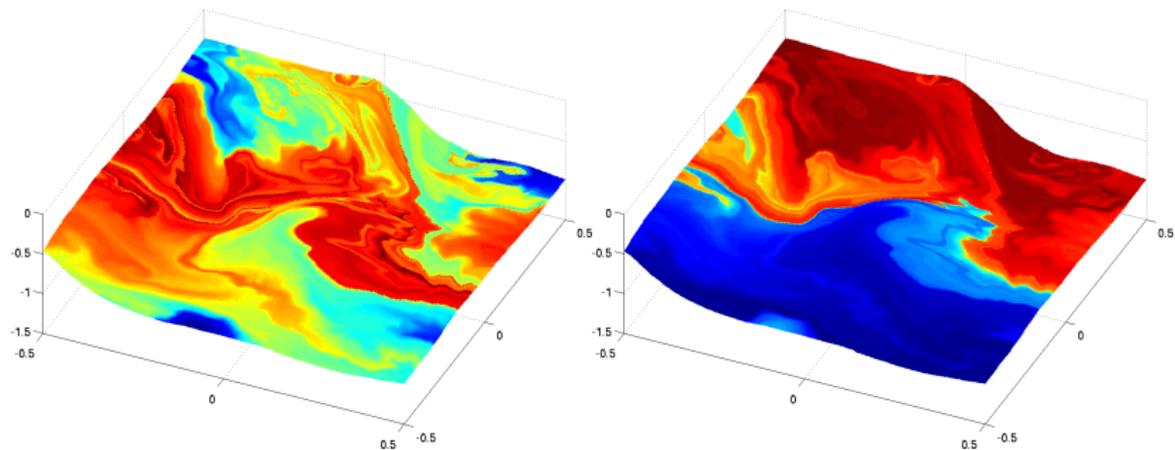
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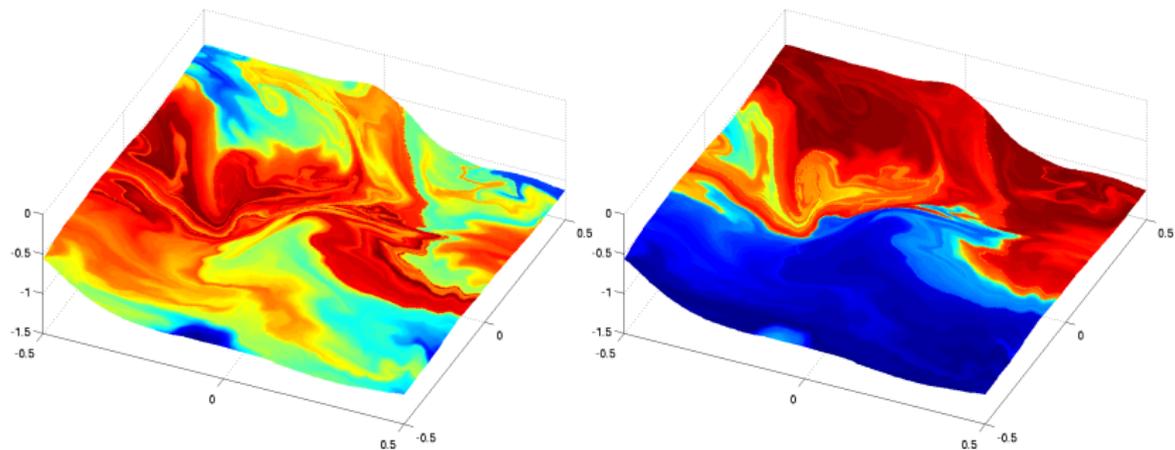
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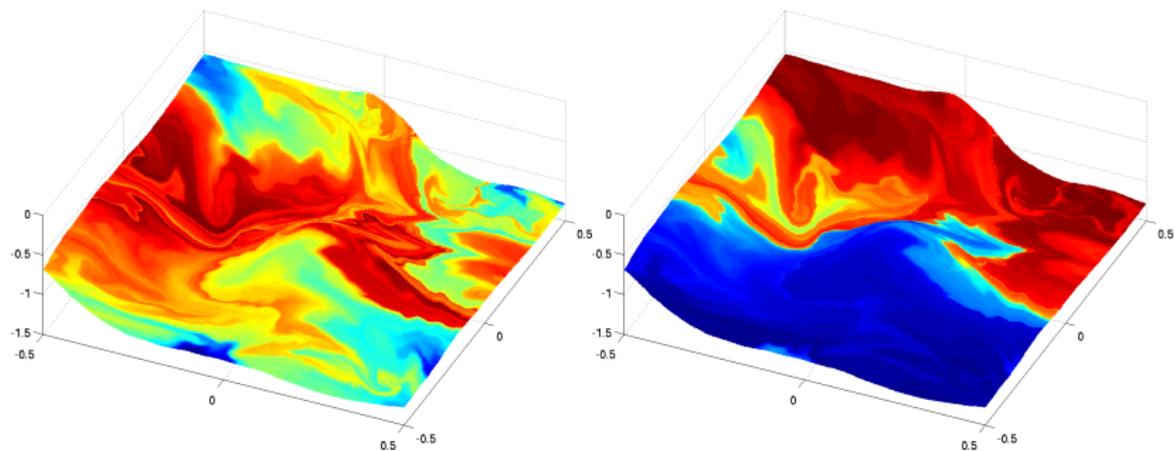
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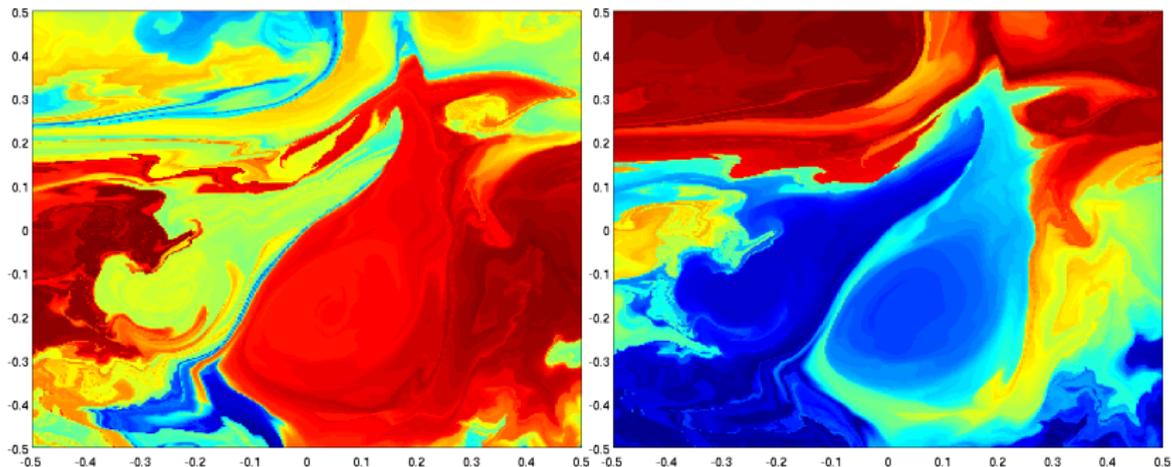
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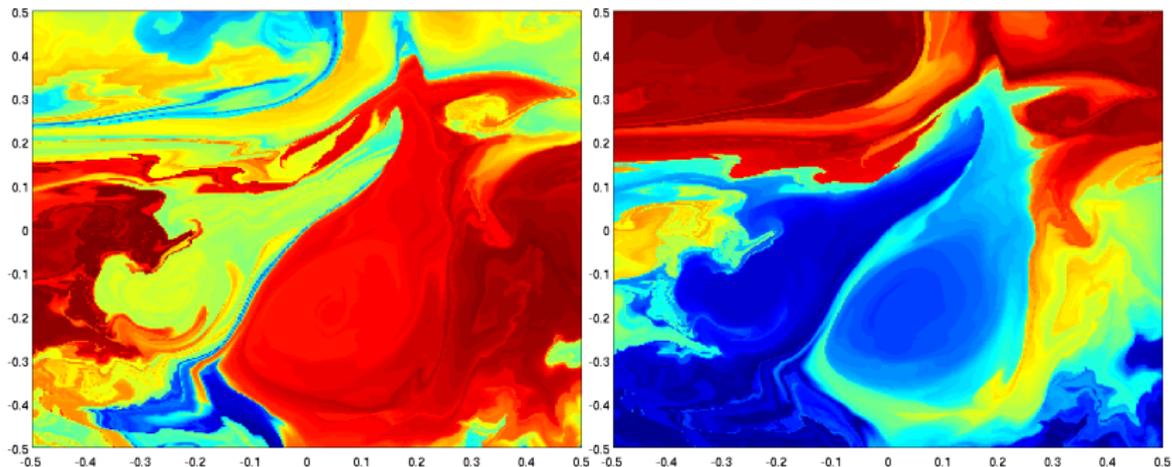
Depth dependence: Observable $k_x = 0$, $k_y = 1/2$



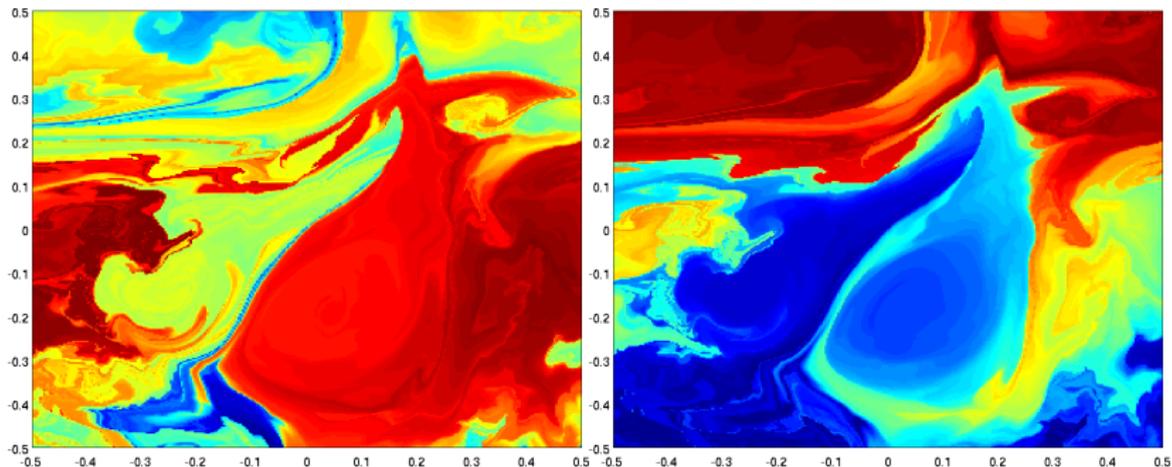
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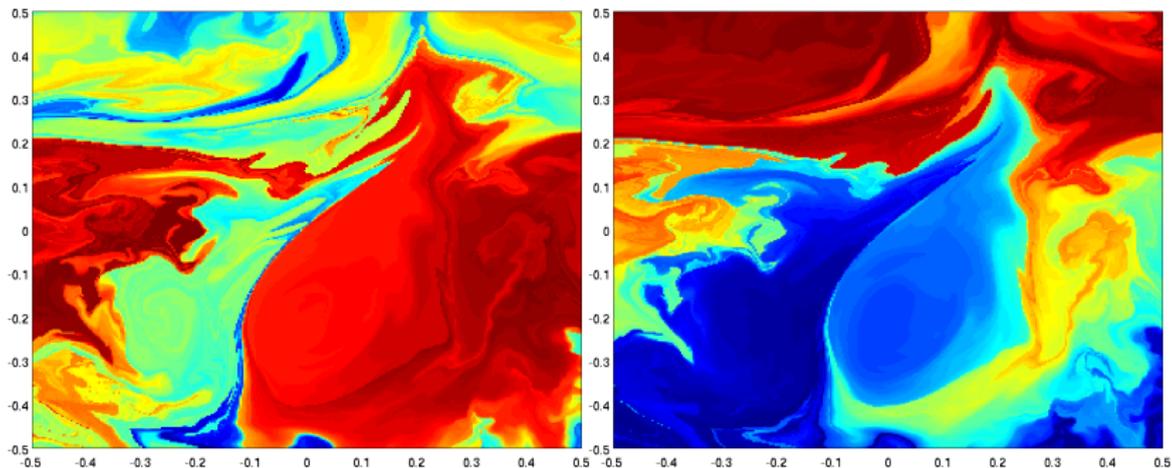
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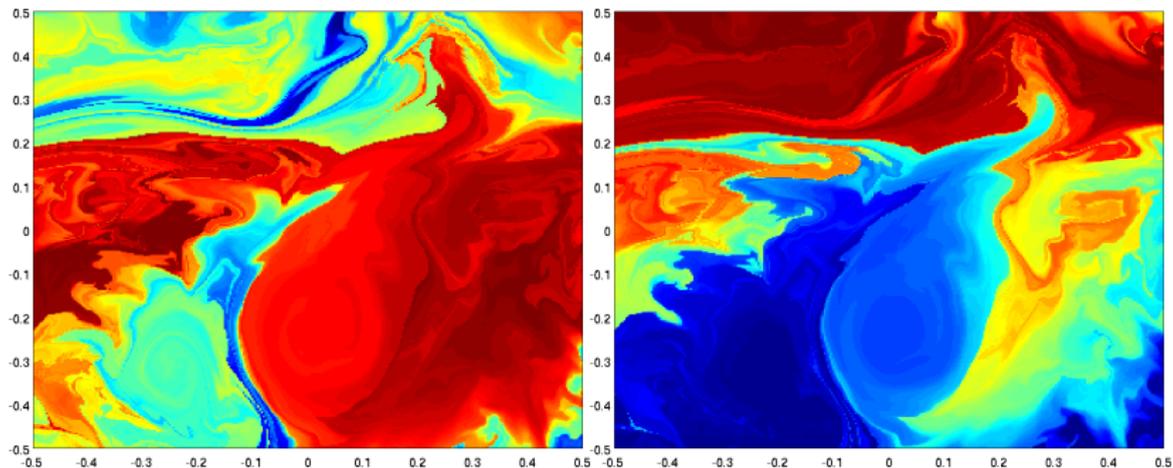
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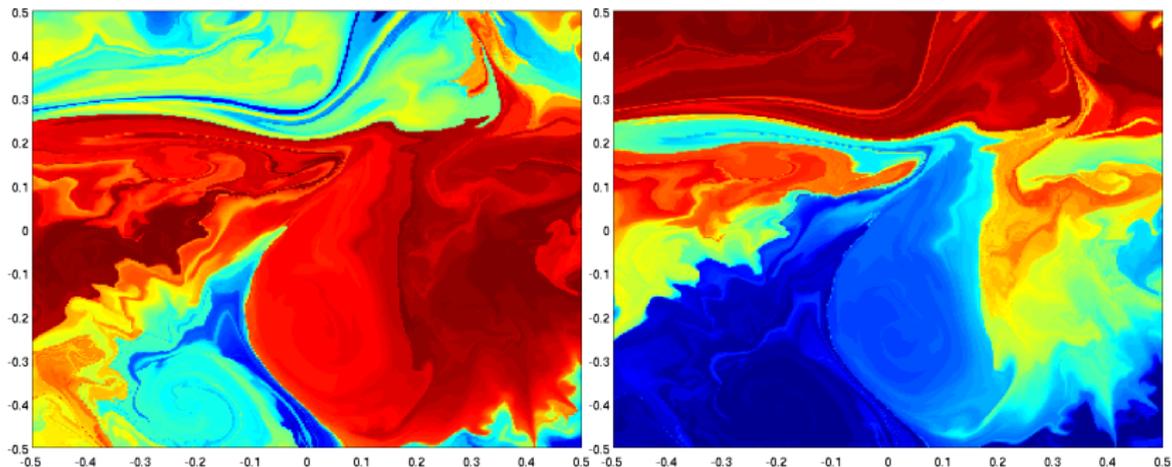
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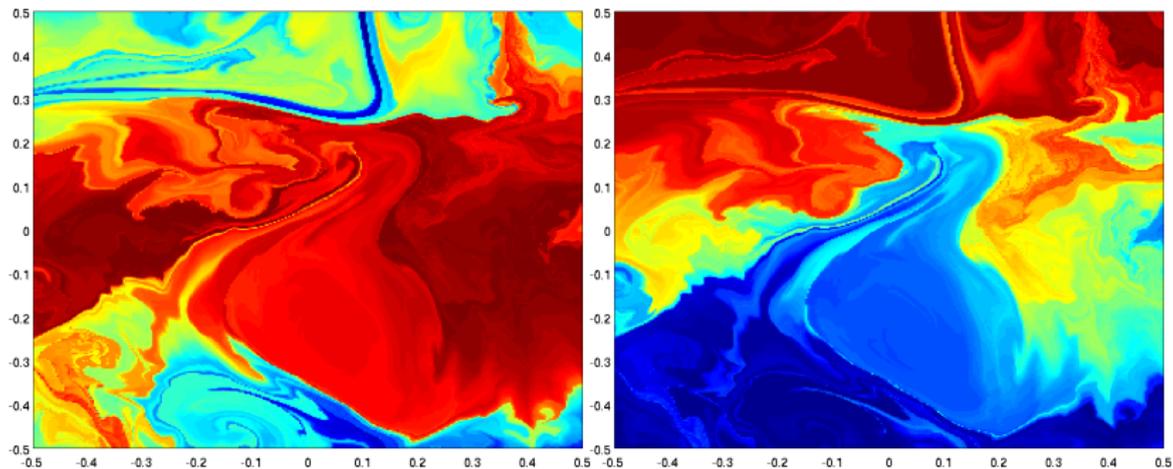
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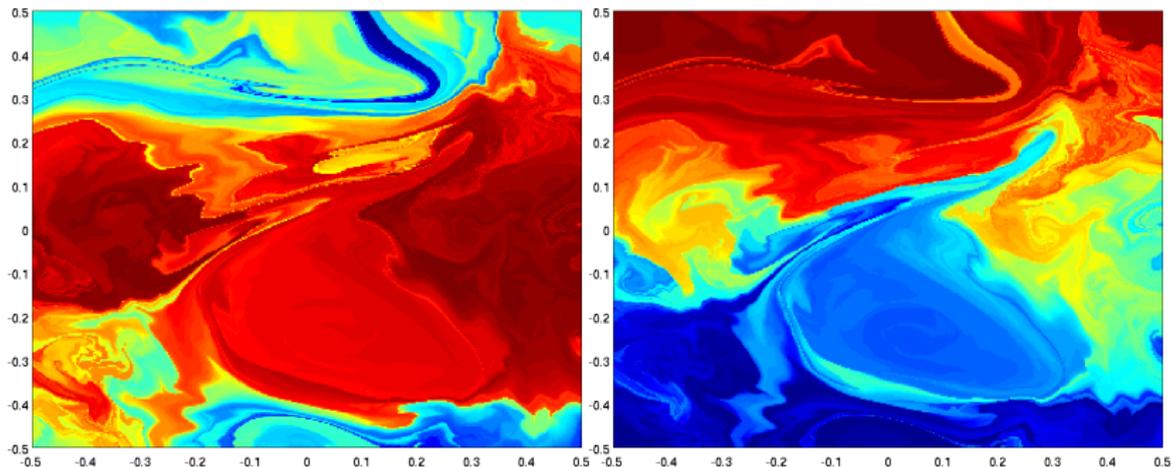
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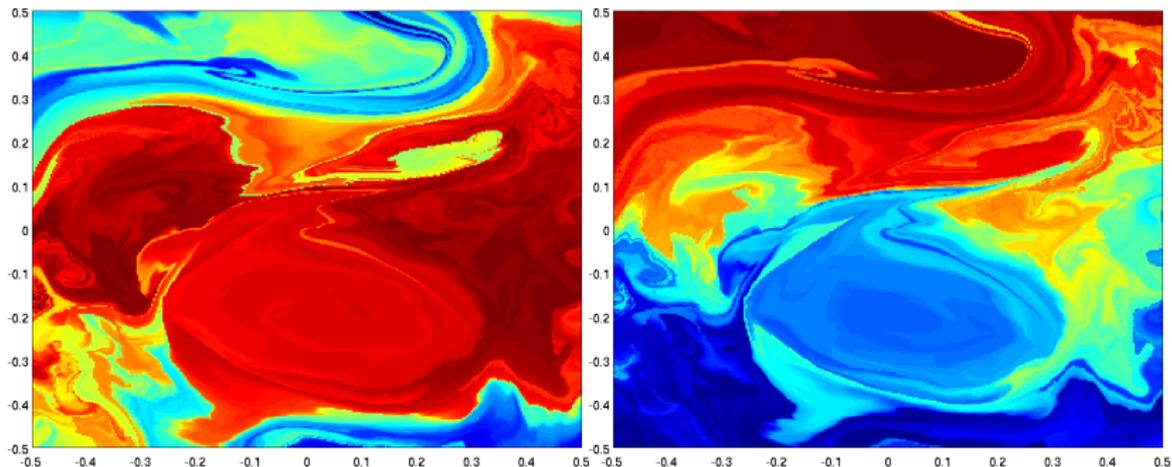
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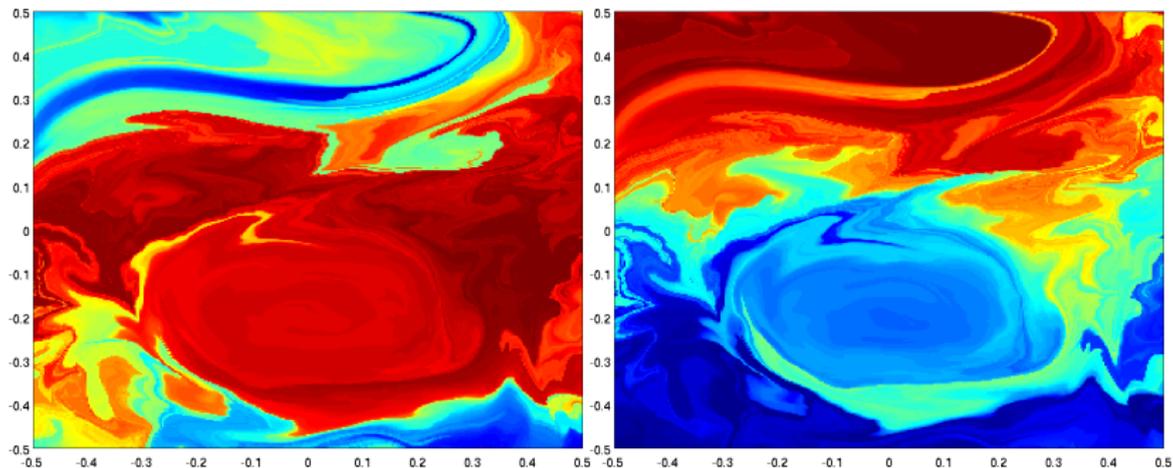
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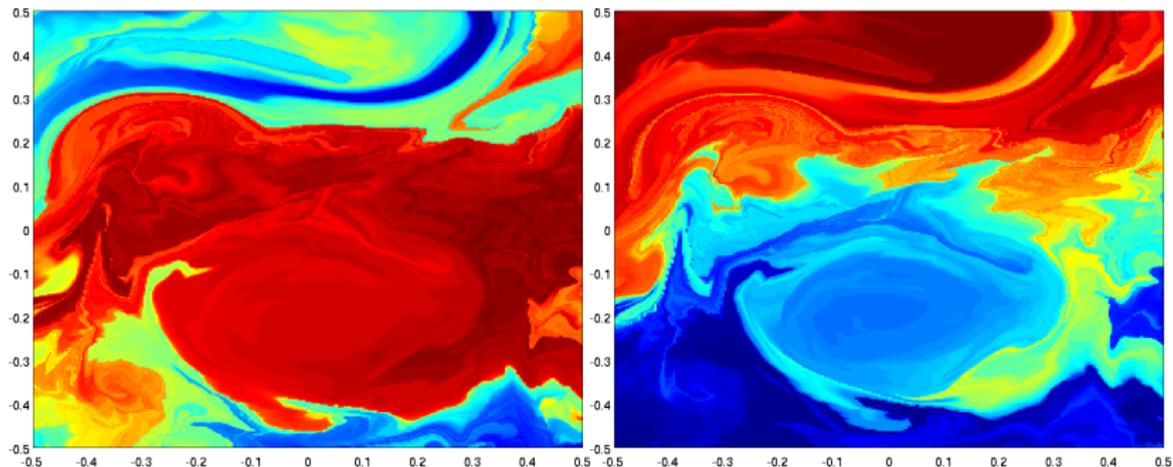
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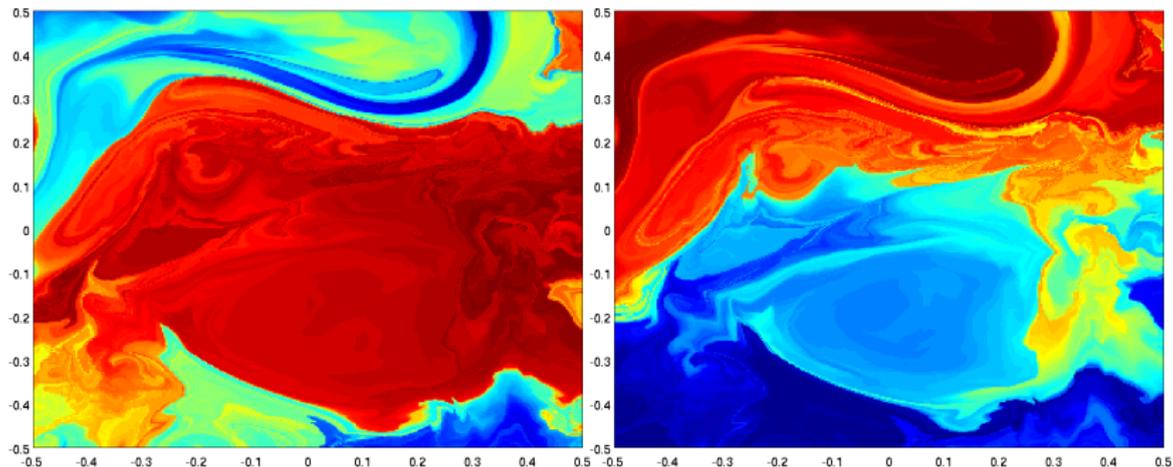
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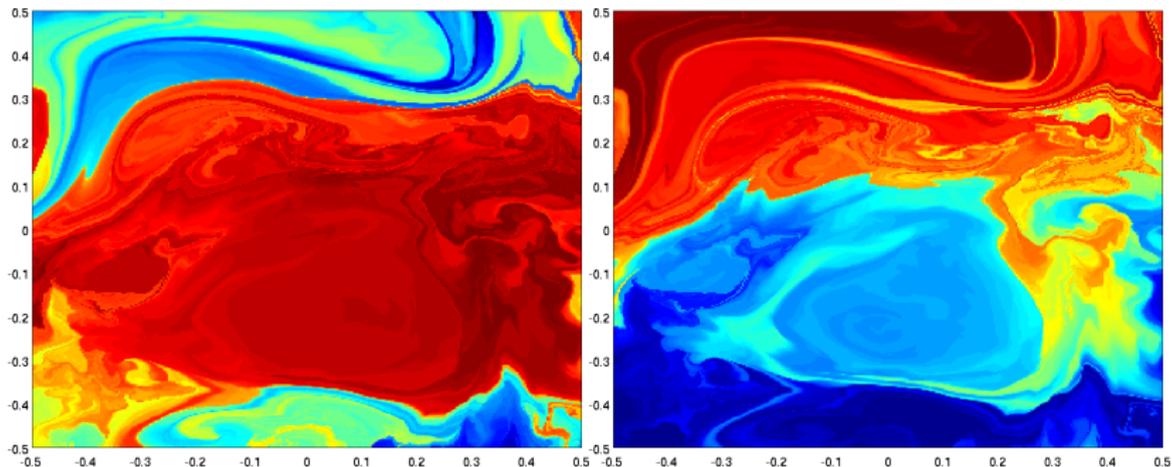
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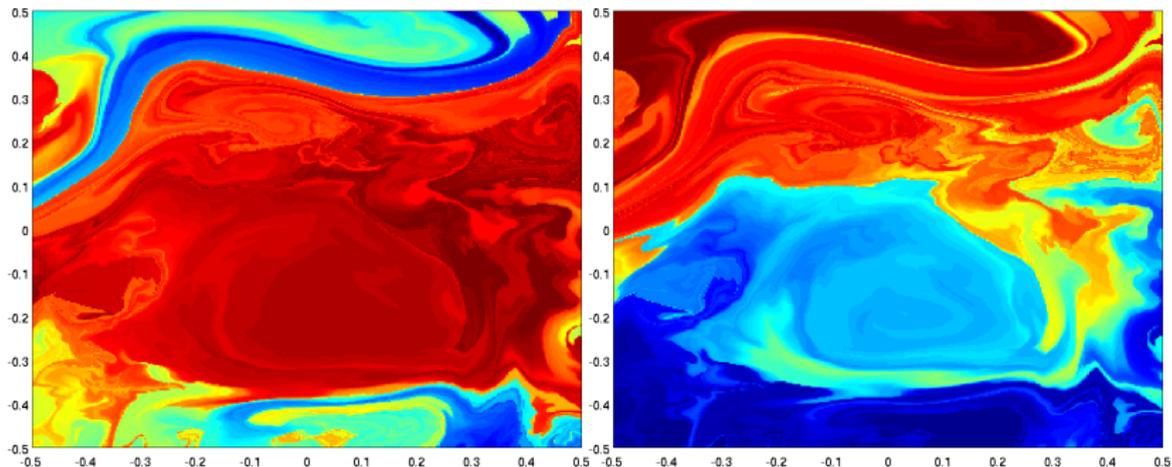
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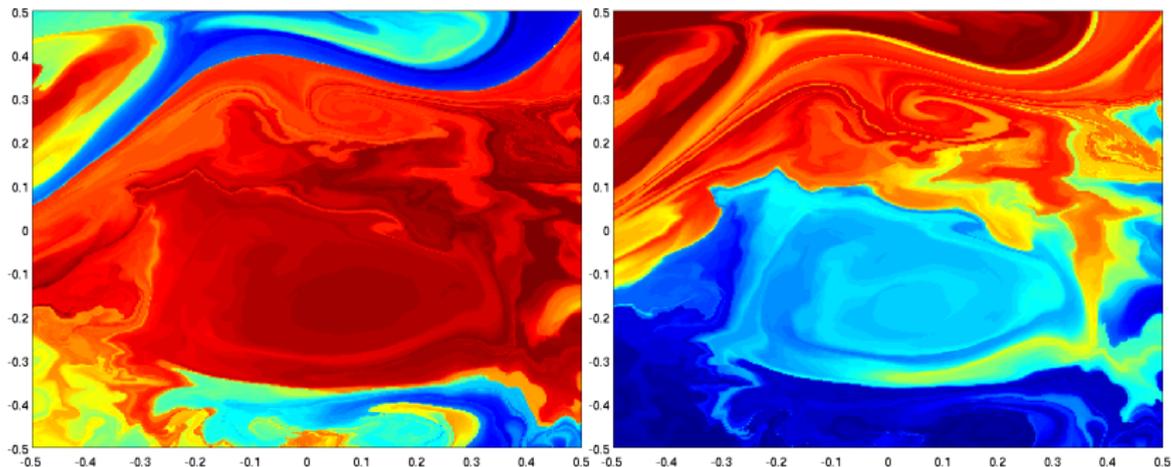
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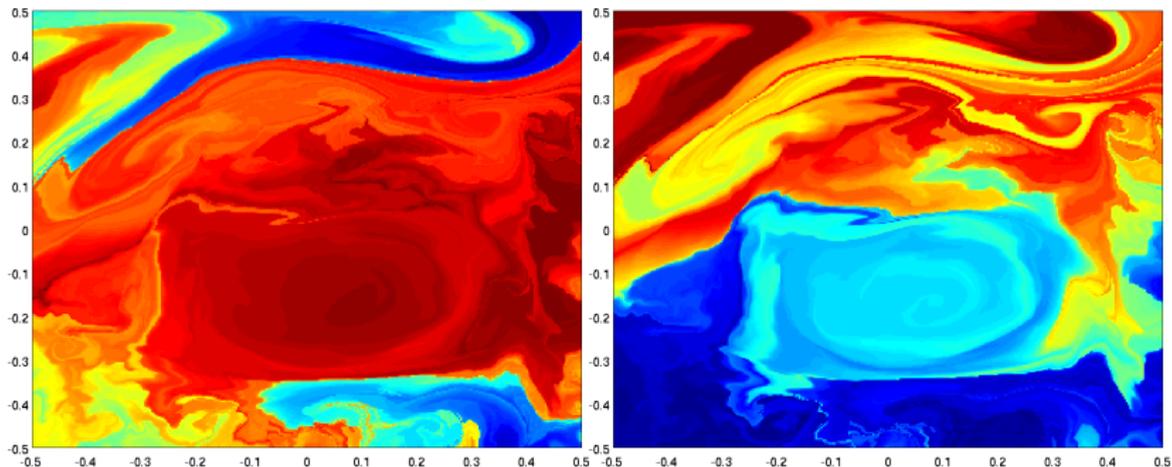
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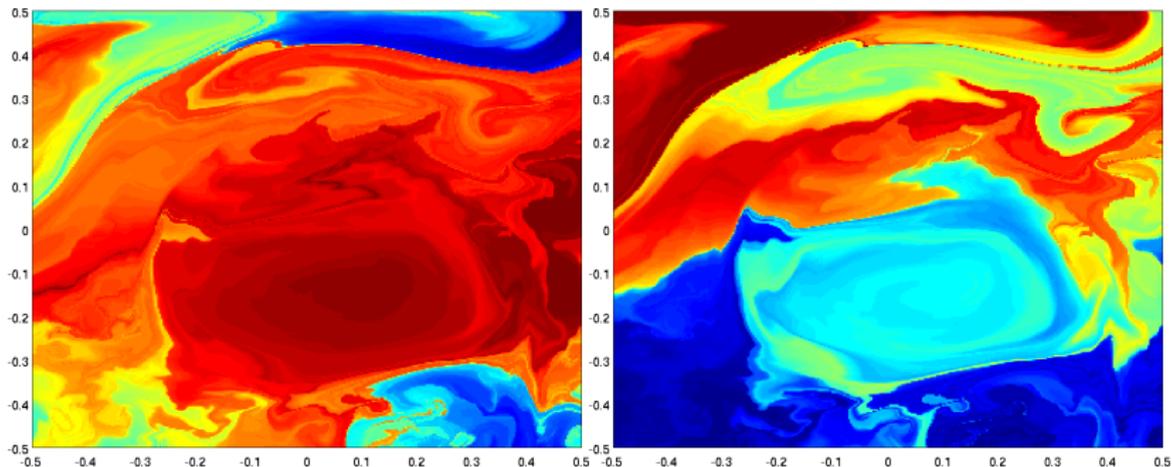
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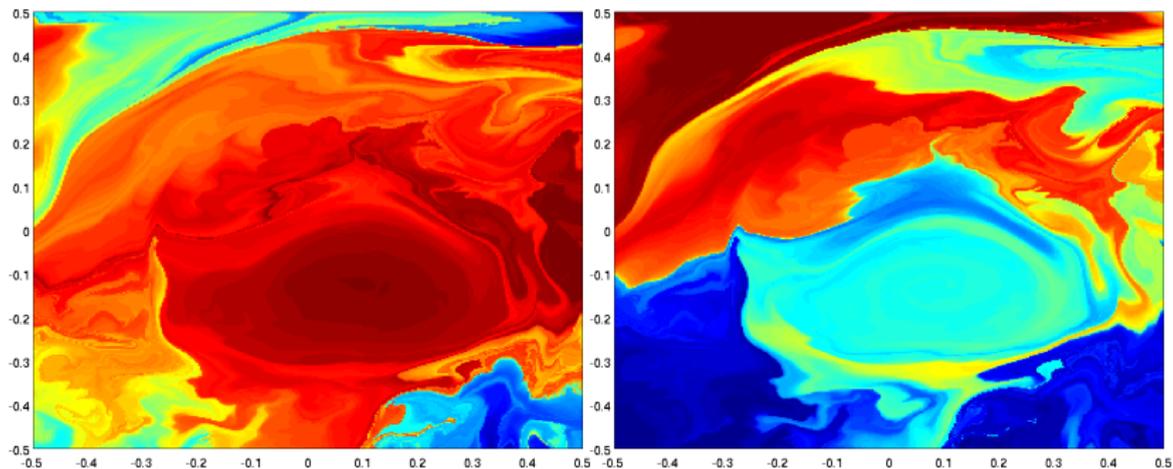
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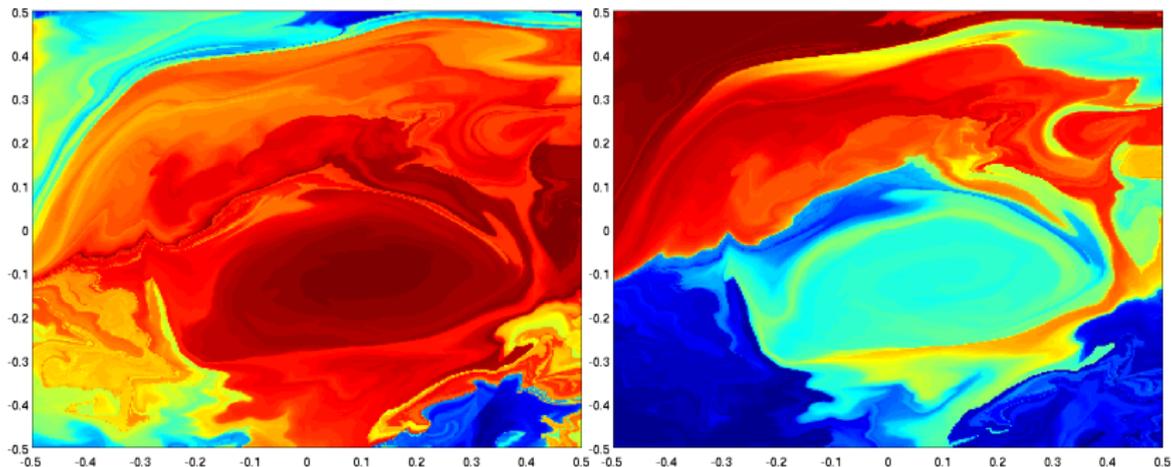
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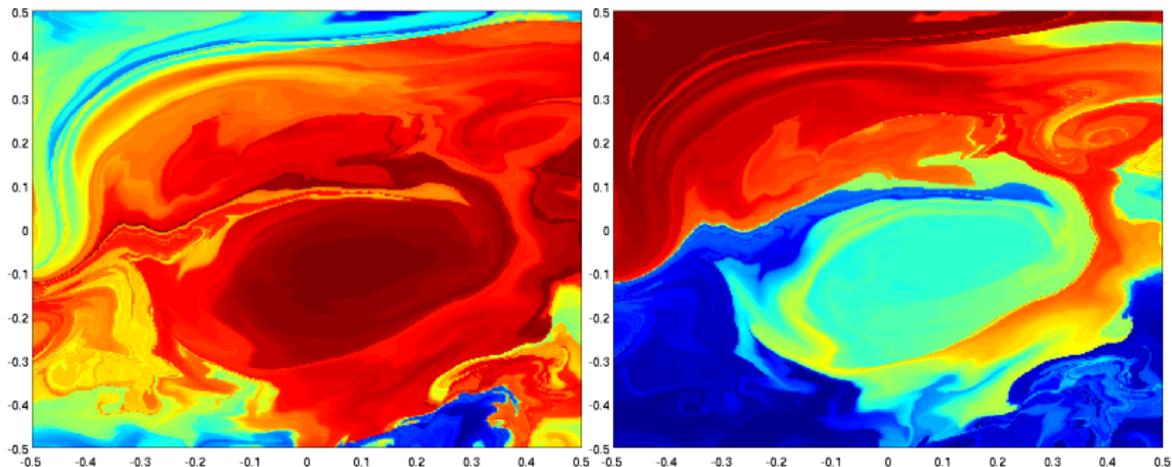
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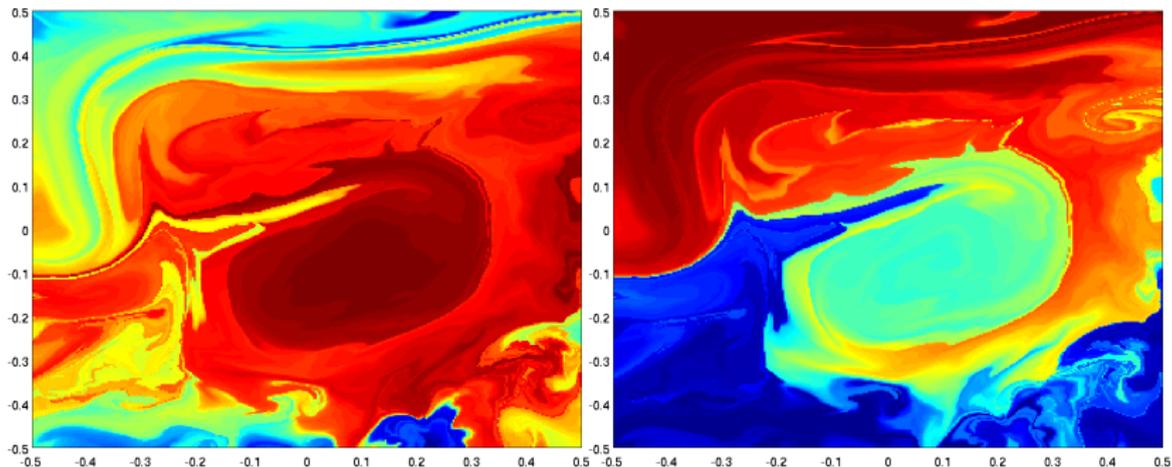
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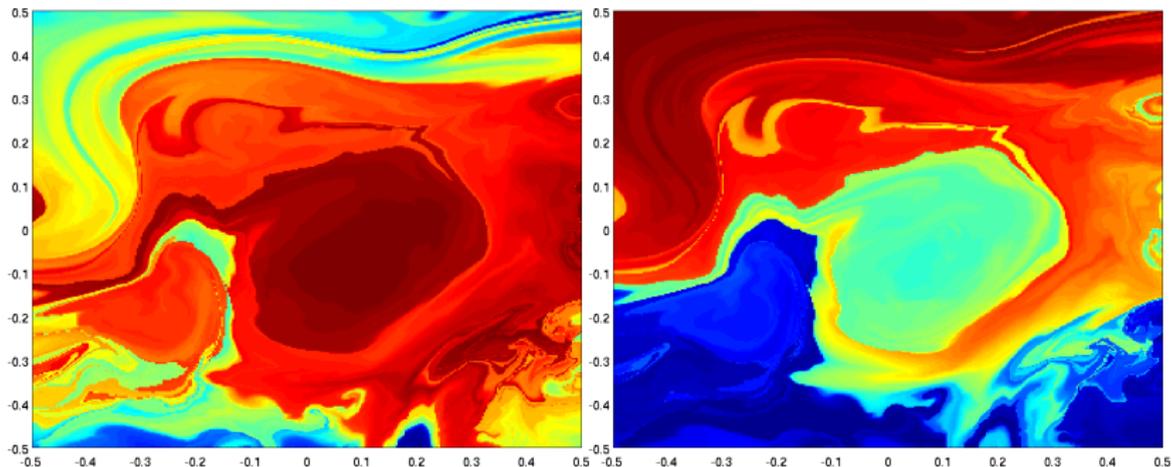
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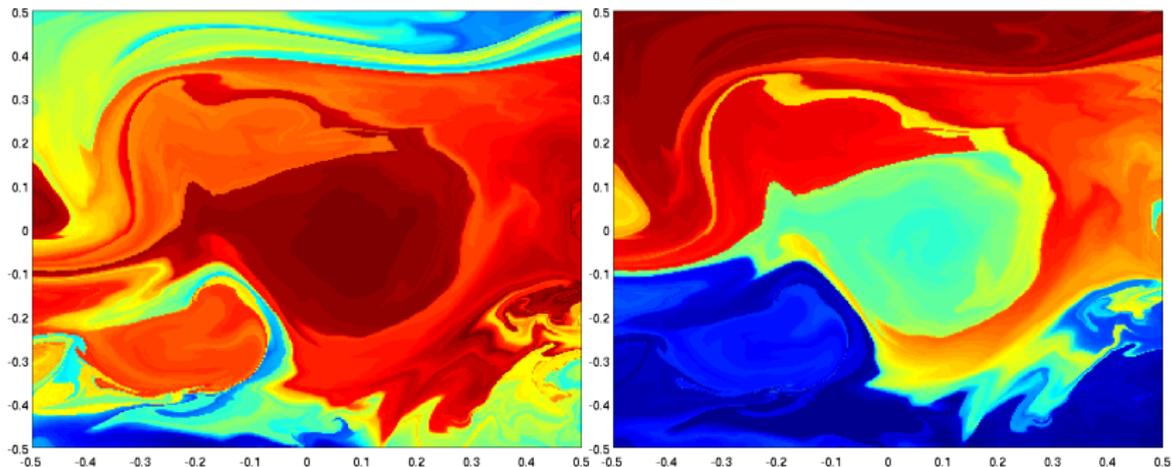
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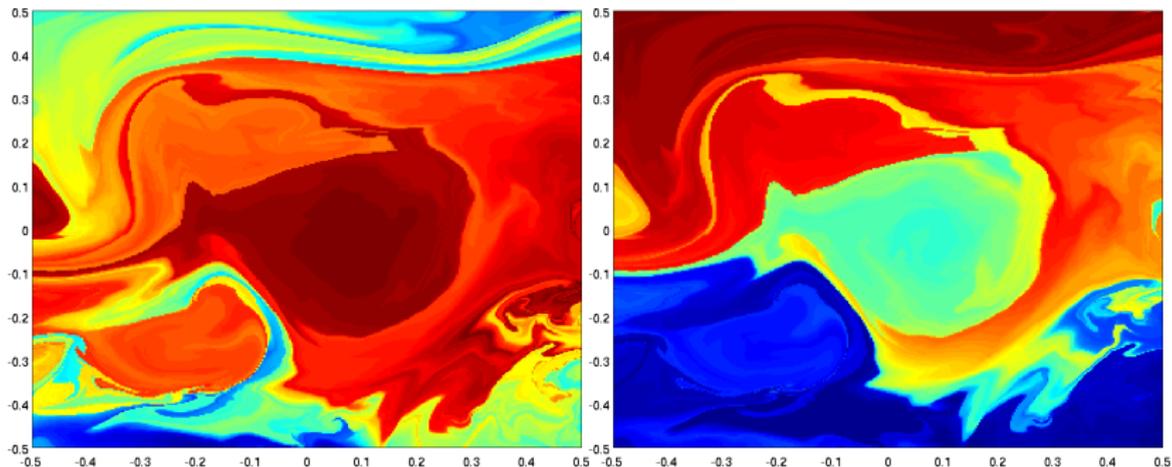
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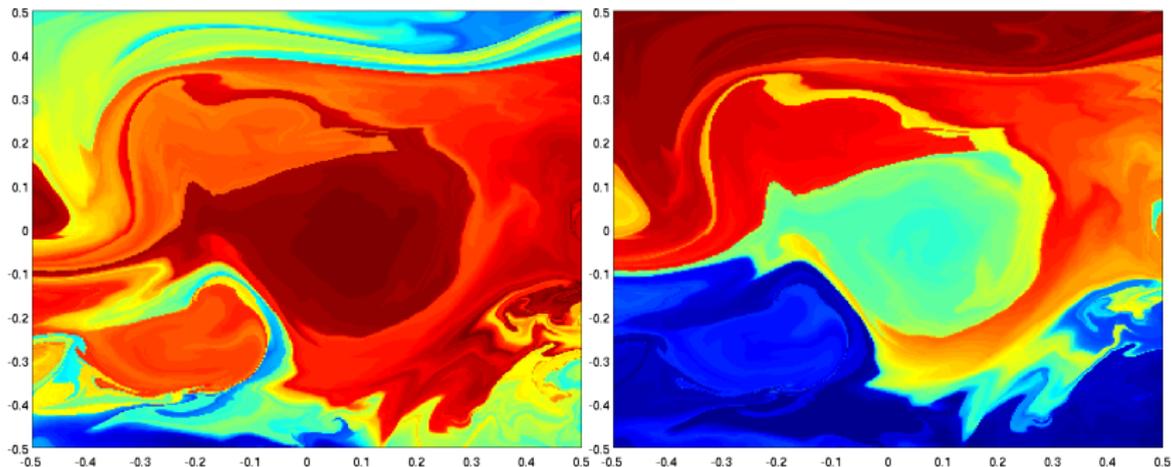
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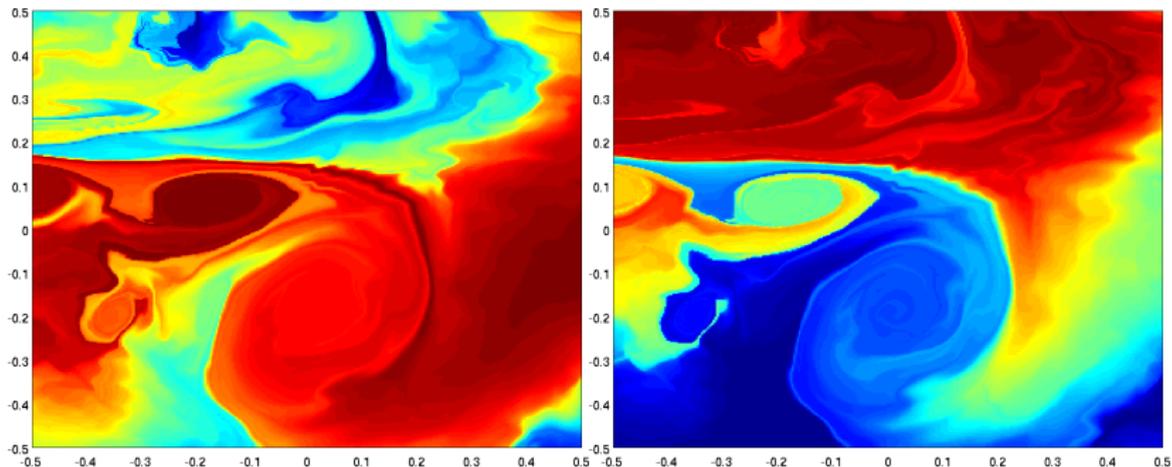
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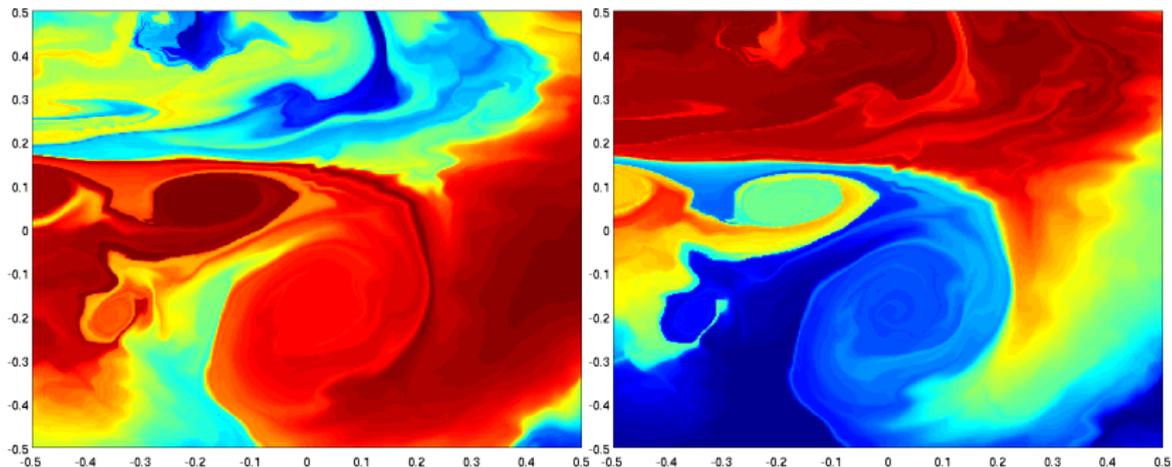
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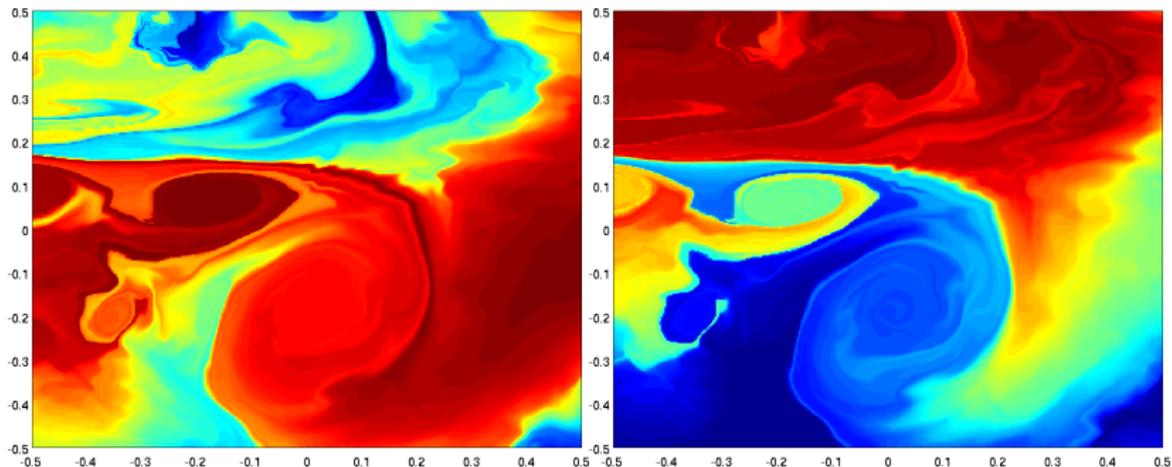
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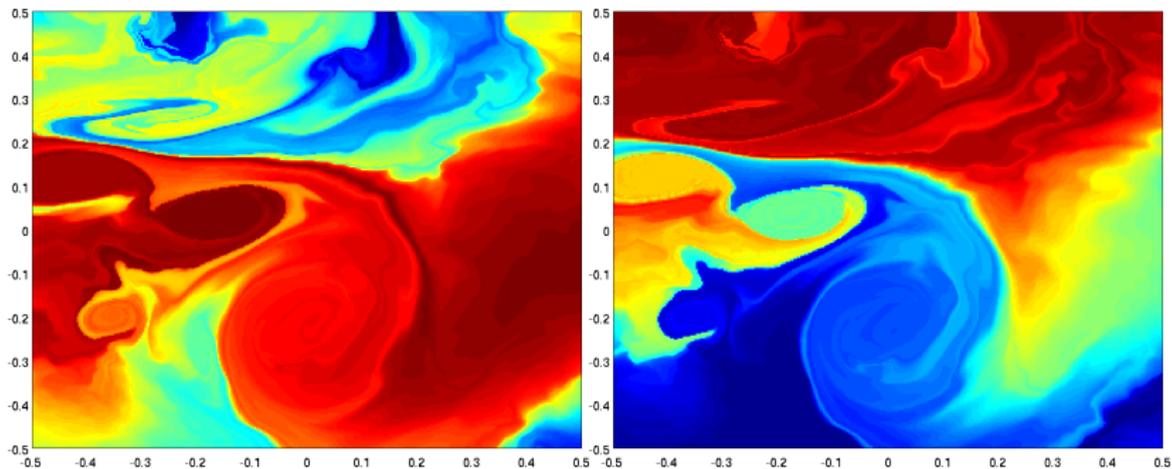
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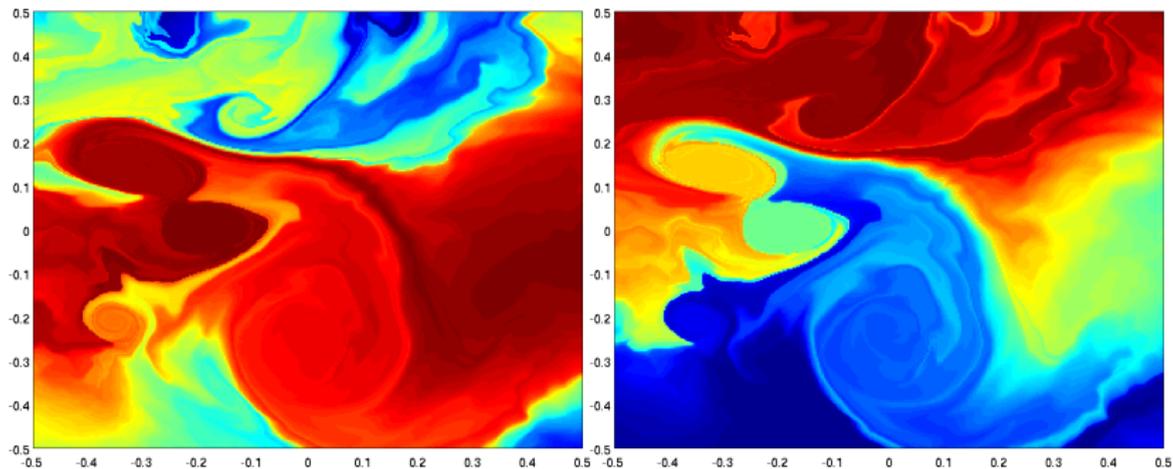
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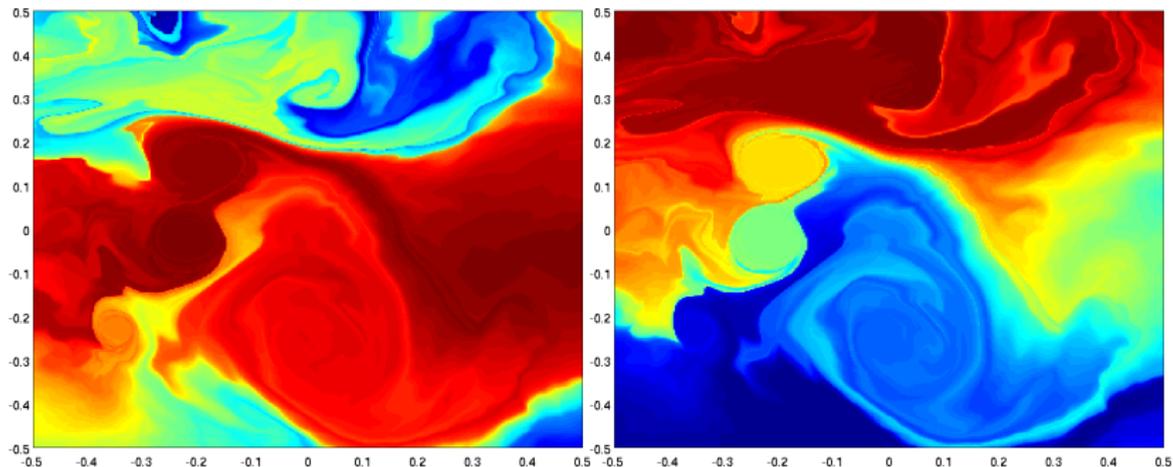
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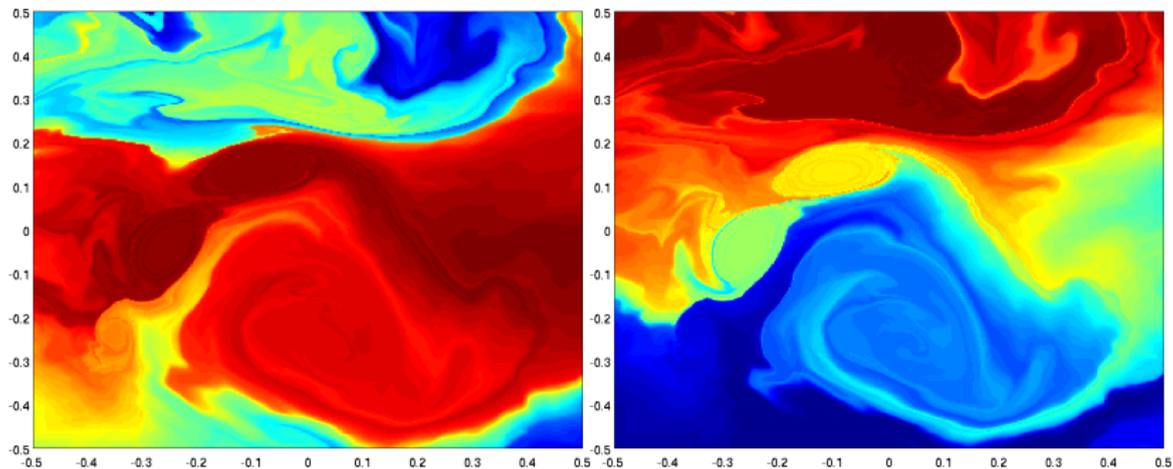
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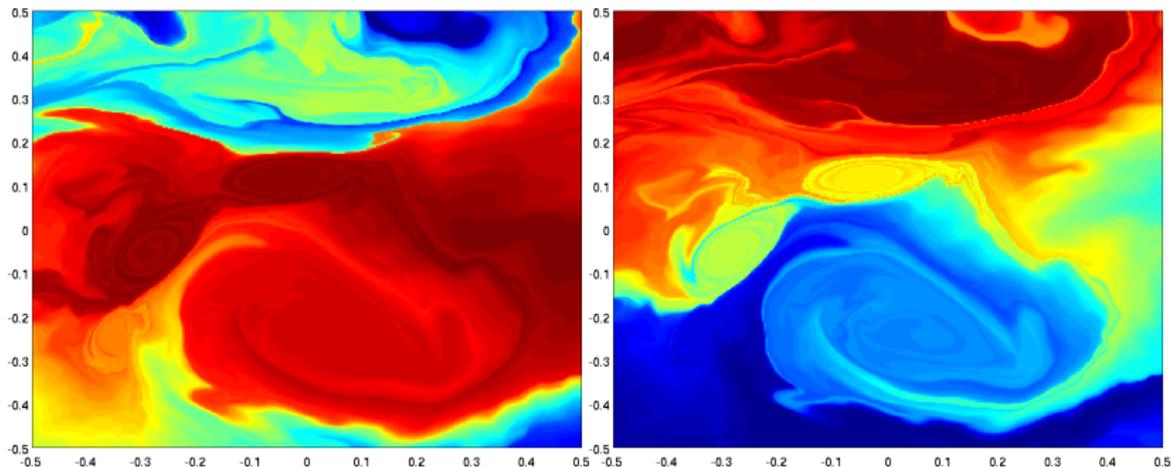
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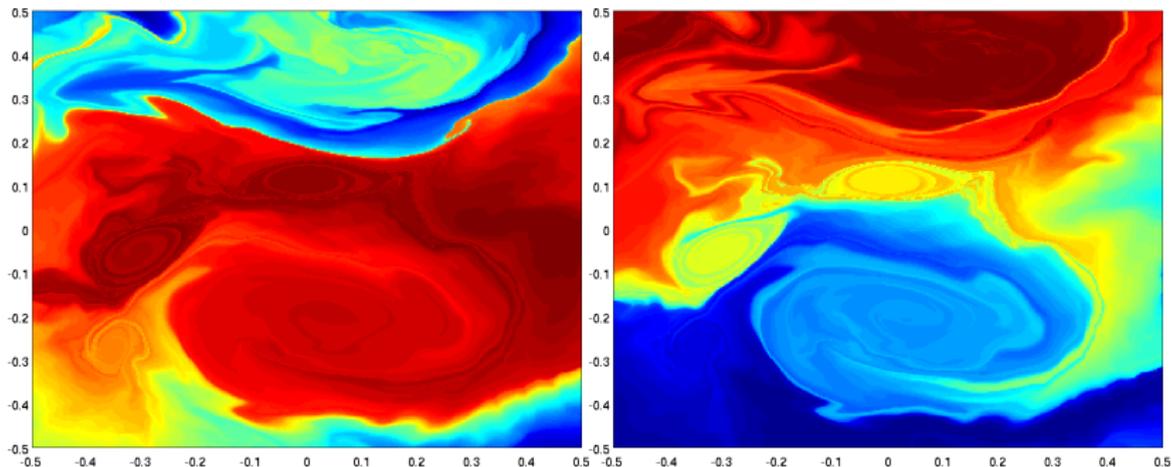
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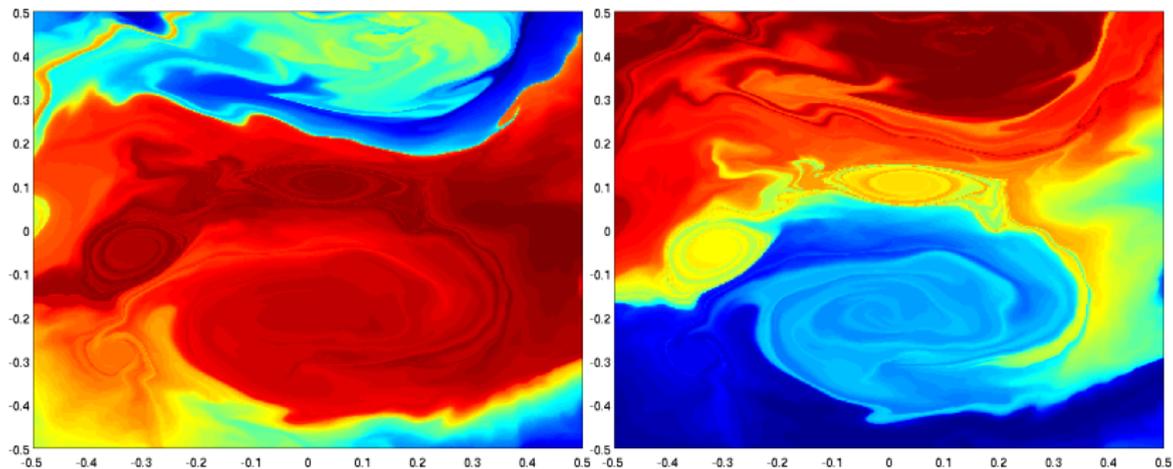
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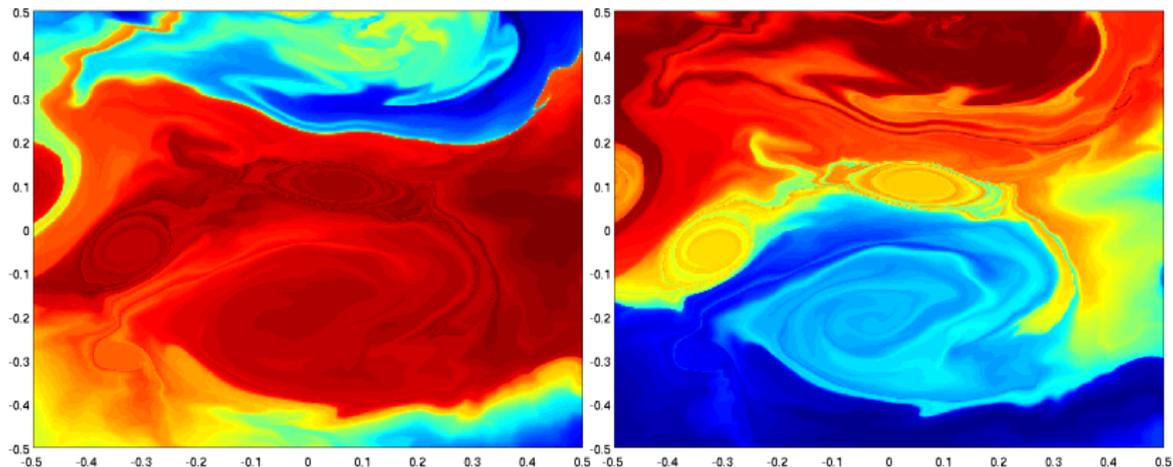
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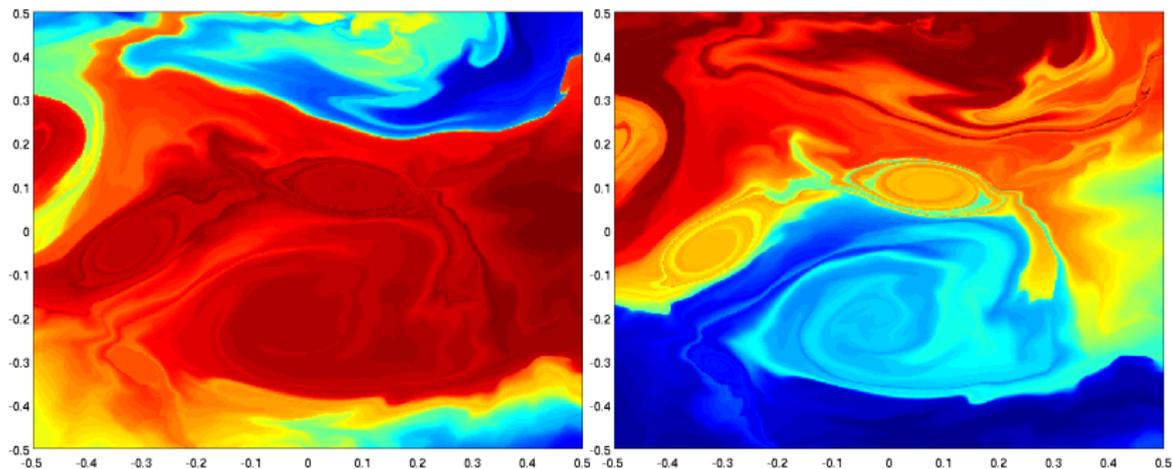
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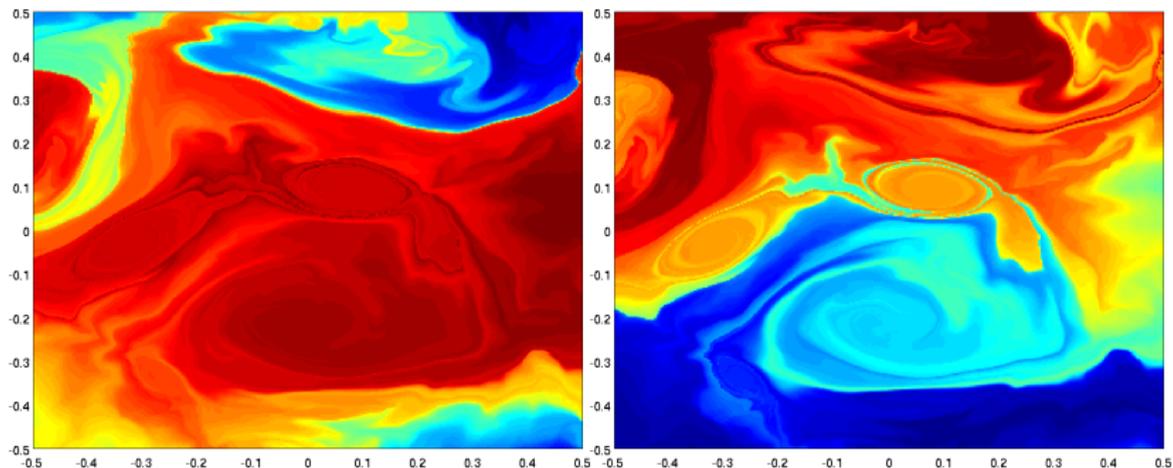
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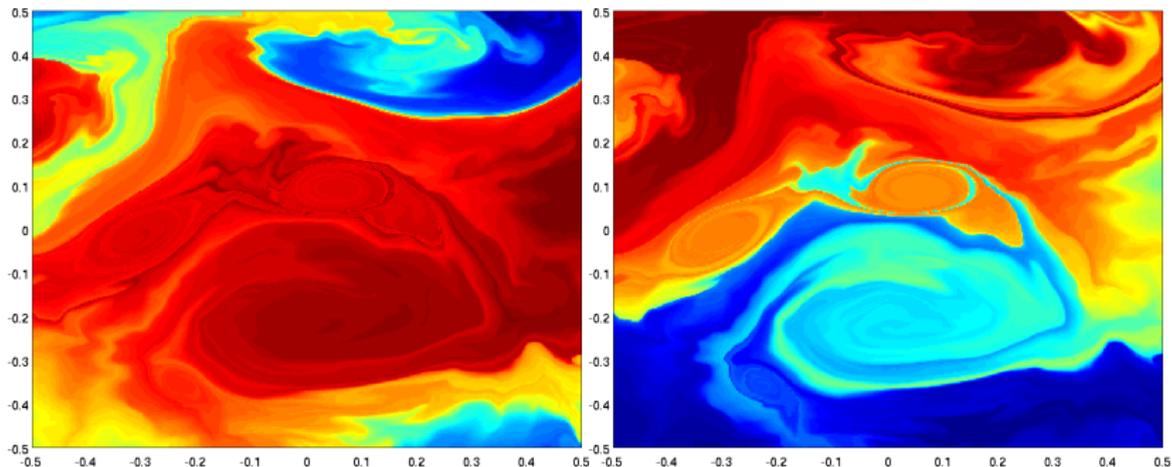
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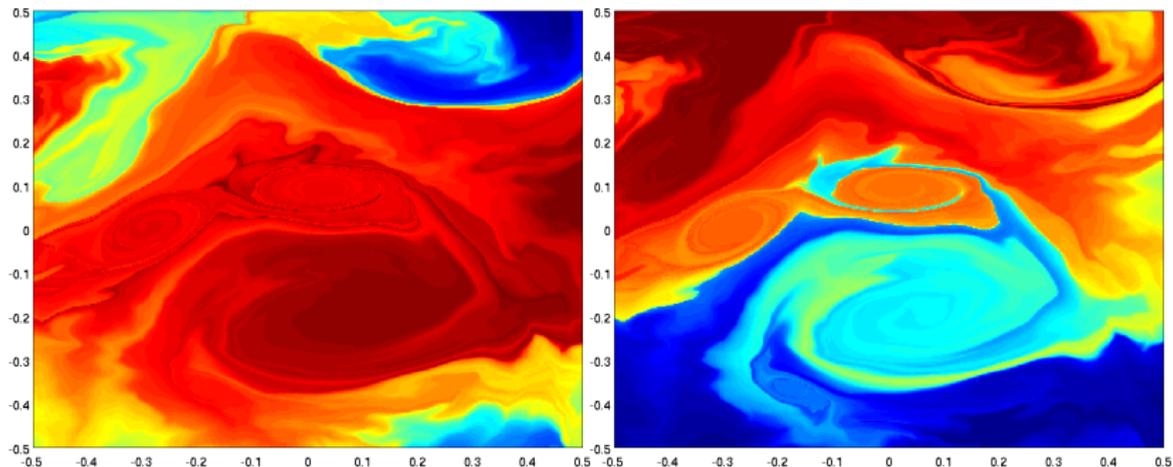
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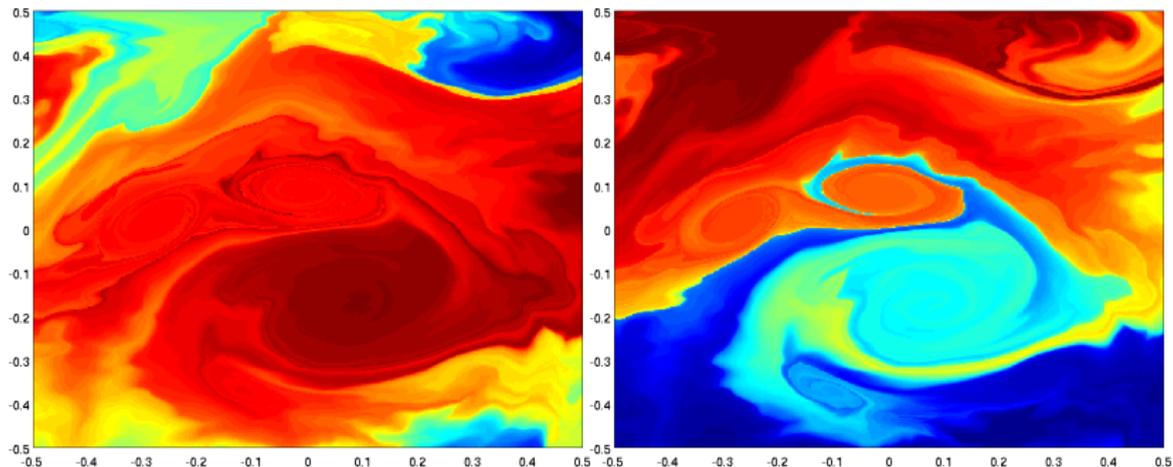
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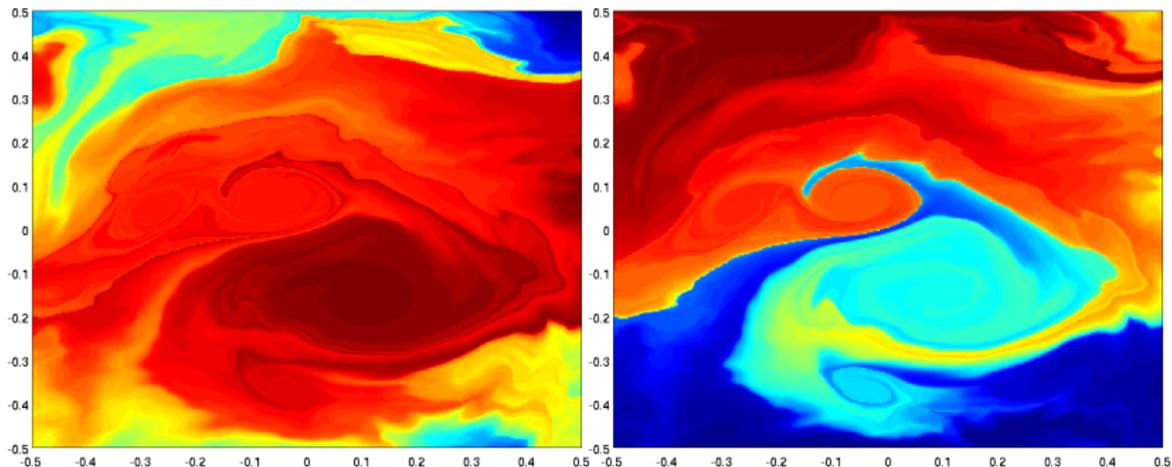
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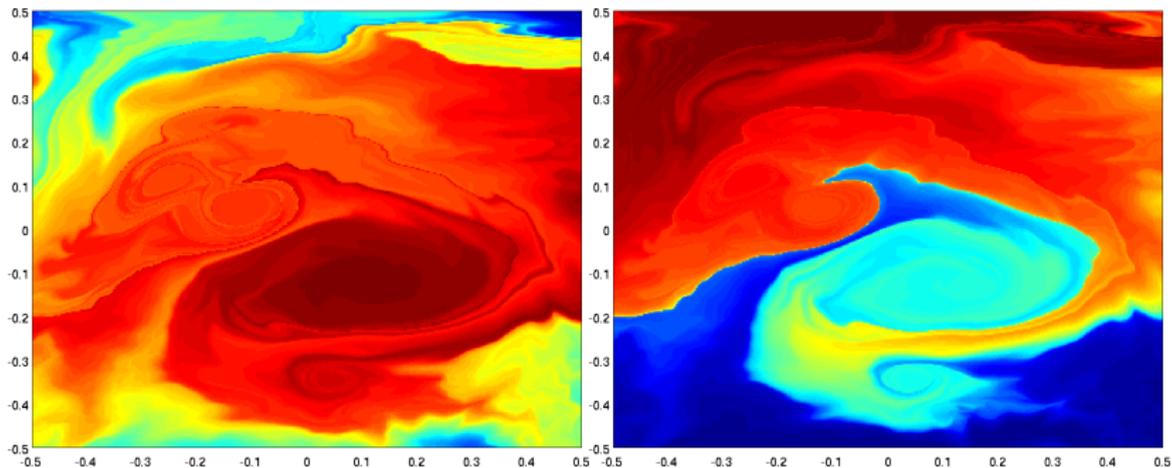
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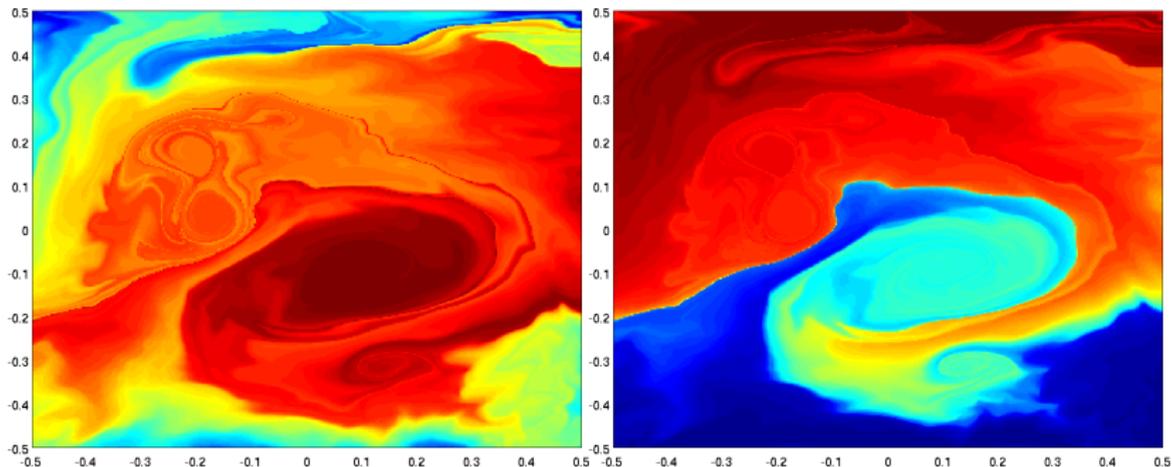
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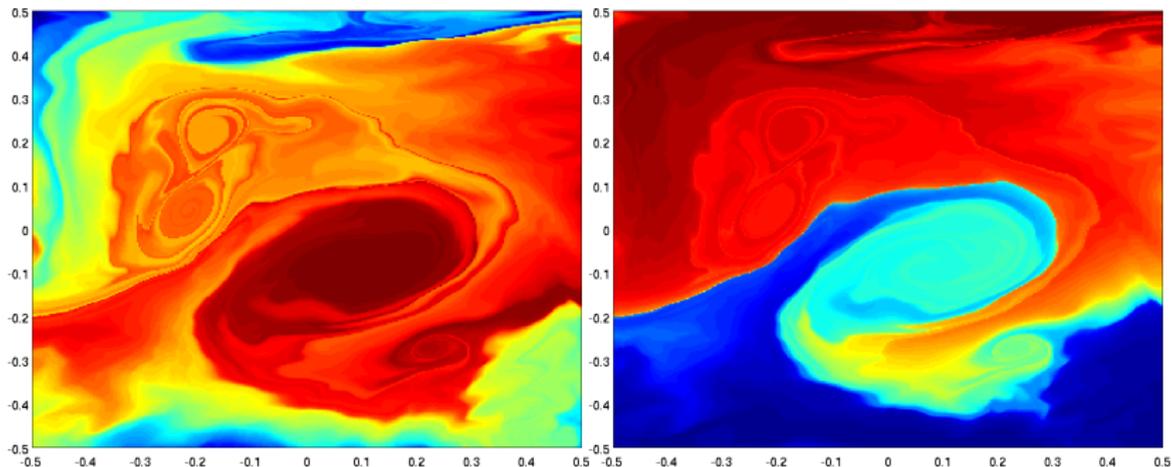
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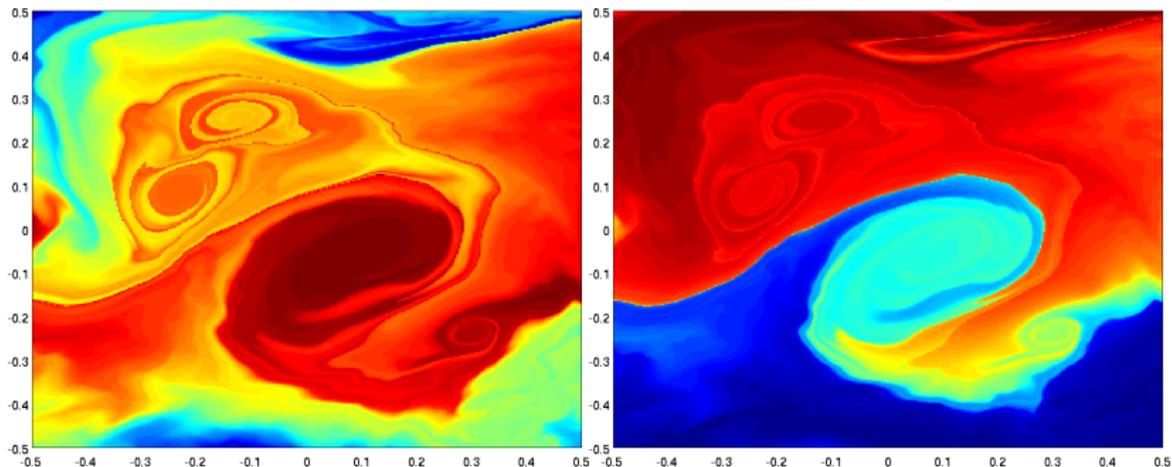
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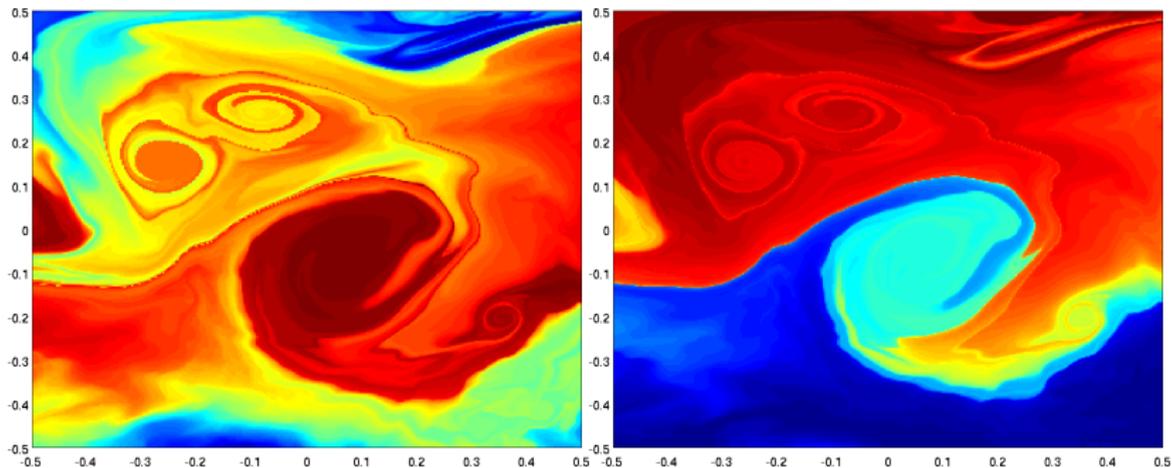
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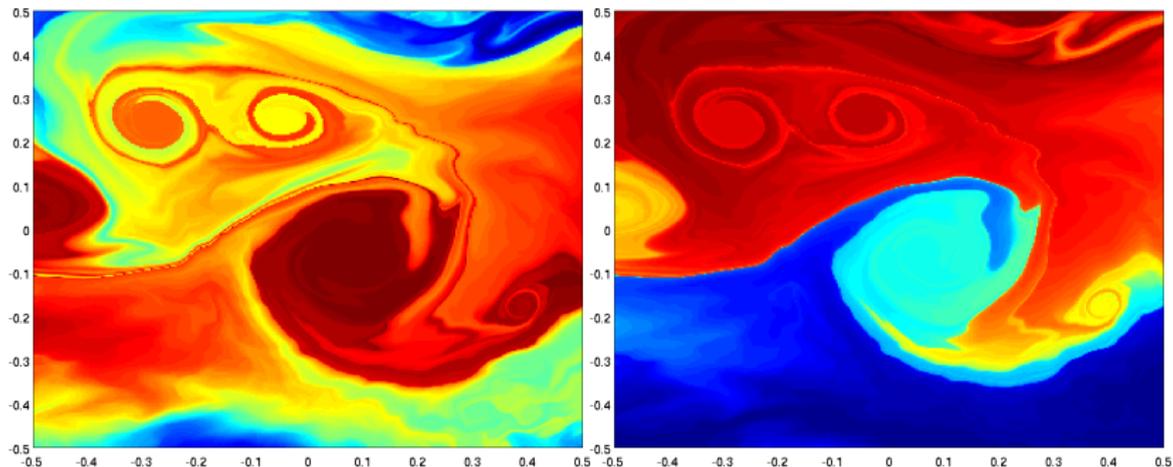
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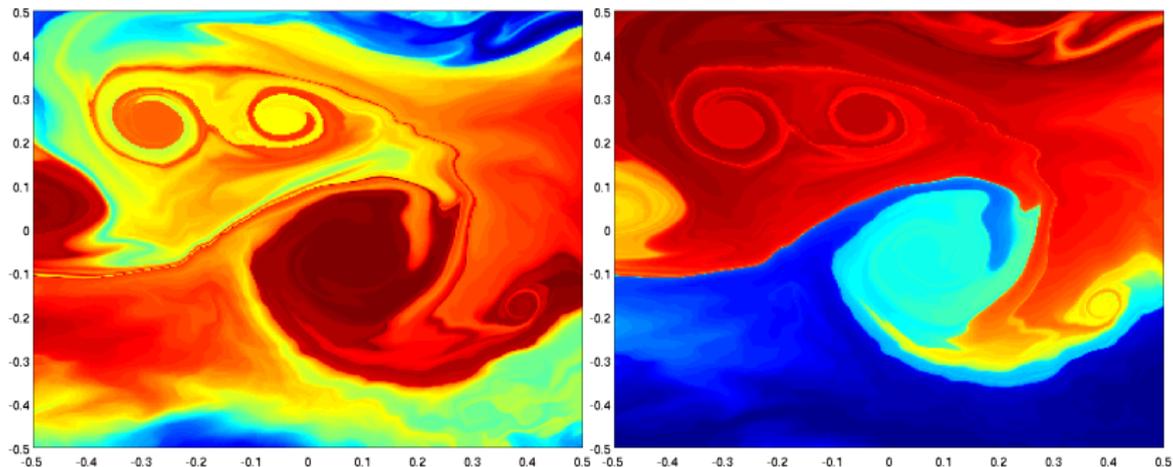
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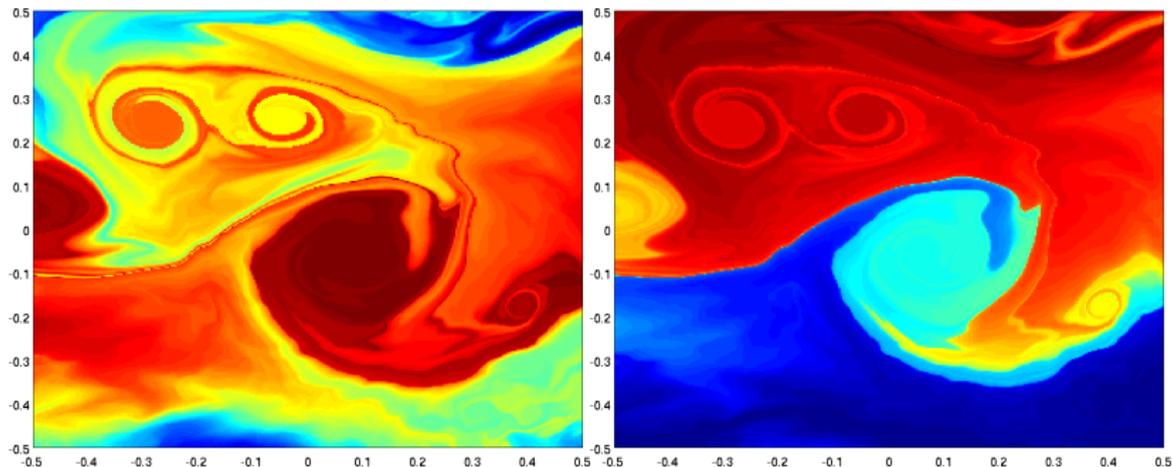
DAY=90



DAY=90



DAY=90

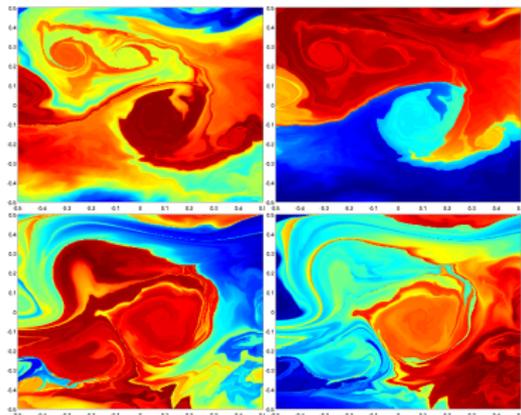


Partition: By Isovalues in \mathbb{R}^n

- Group by iso-values of single observable, $f_k(\mathbf{x}_0)$
- Different kinematics - similar isovalues.
- Consider simultaneous vector of observables:

$$\mathbf{f}(\mathbf{x}) = \{f_1(\mathbf{x}), f_2(\mathbf{x}), \dots, f_n(\mathbf{x})\}$$

- On averaging: $\bar{\mathbf{f}} : \mathbf{x}_0 \rightarrow \mathbb{R}^n$
- Color \mathbf{x}_0 by similarity in \mathbb{R}^n (cluster algorithm)

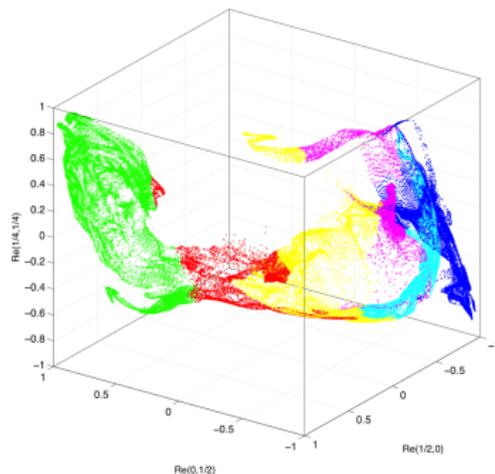


A Very Simple Example:

- Basis Set:

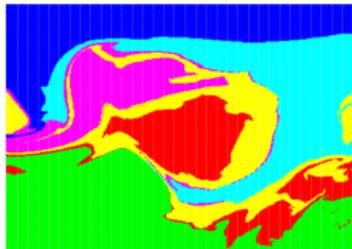
$$f_{k,l} = \cos(kx) \cos(l y) + i \sin(kx) \sin(l y)$$

- $k = \{0, 1/4, 1/2\}$
- $(9 \times 10^4) \times 23$ points in \mathbb{R}^{16}
- Standard, quick, (dumb?)
k-means clustering
- 6 clusters

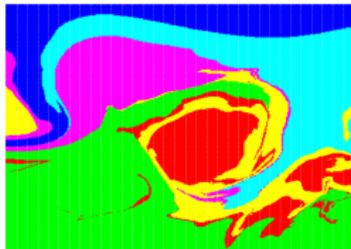


6 Cluster Partition: Day 90

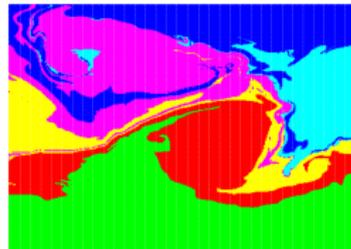
Isopycnal 1



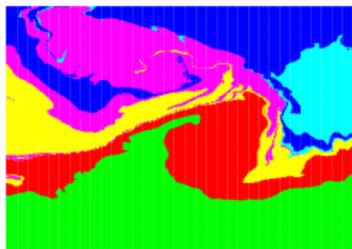
Isopycnal 7



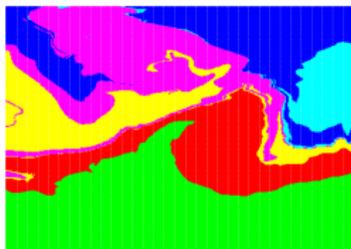
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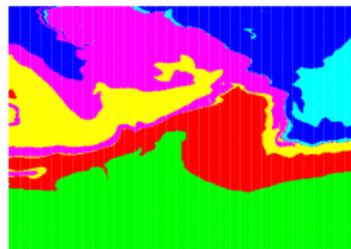
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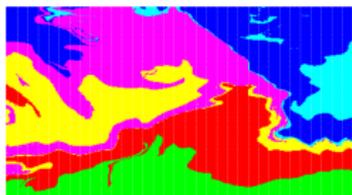
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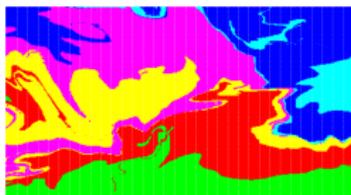
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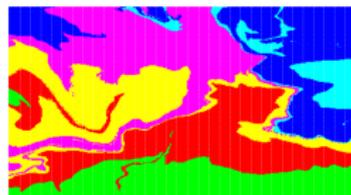
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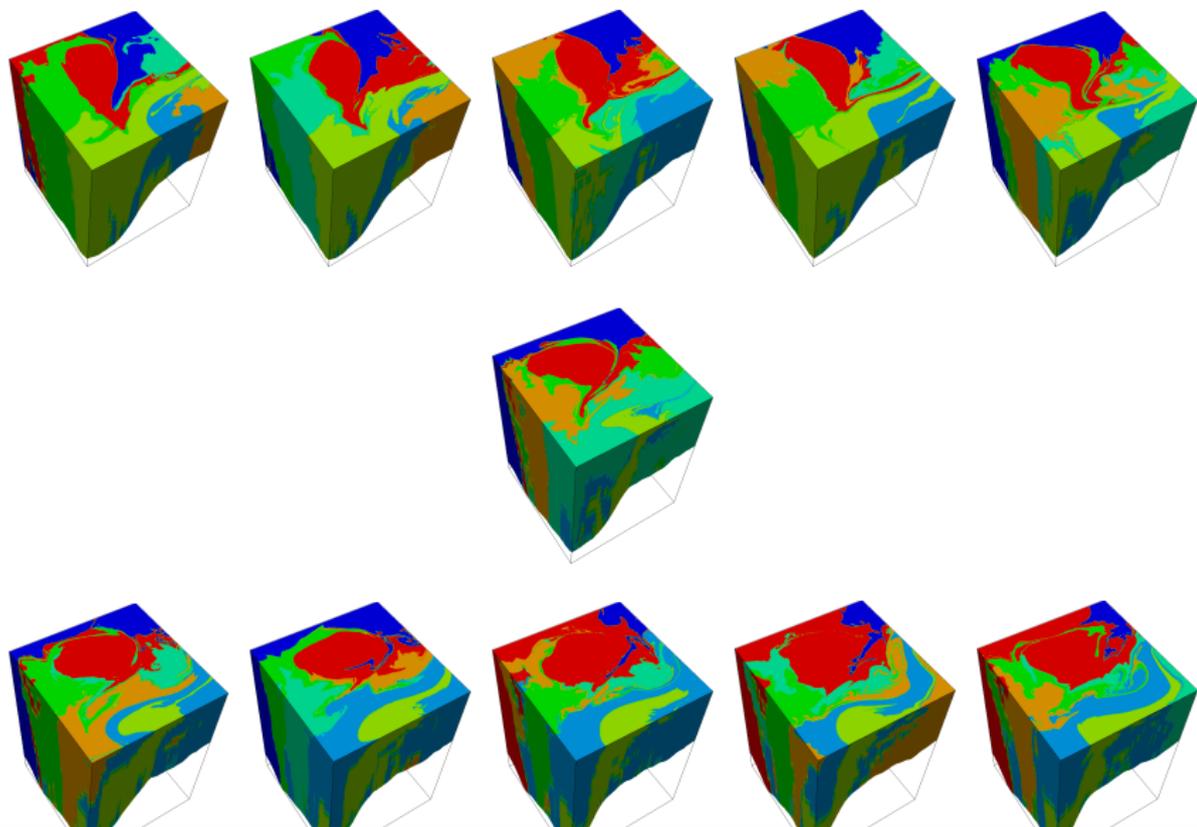
Isopycnal 22



Isopycnal 23



Partition: 3D Time Dependence



Summary:

- *Koopman* approach - simple, readily implemented, efficient means of identifying advective transport geometry in Eulerian frame.
 - ▶ Select: τ and nominal scales of interest through \mathbf{f} .
 - ▶ Arbitrary dimensions. (2D/3D + 1)
 - ▶ Complex flows (arbitrary time/space dependence)
- Open question: Partition in open flows
 - ▶ Optimal basis sets? (Wavelets, ...)
 - ▶ Convergence? Within basis, across bases, with τ ?
 - ▶ Finite-Time: Harmonic (phase) averaging.