

~200 km
↔

Transport by Coherent Lagrangian Vortices

Ryan Abernathey, Columbia/LDEO
George Haller, ETH

MUR L4 SST

Two Paradigms of “Eddy Flux”

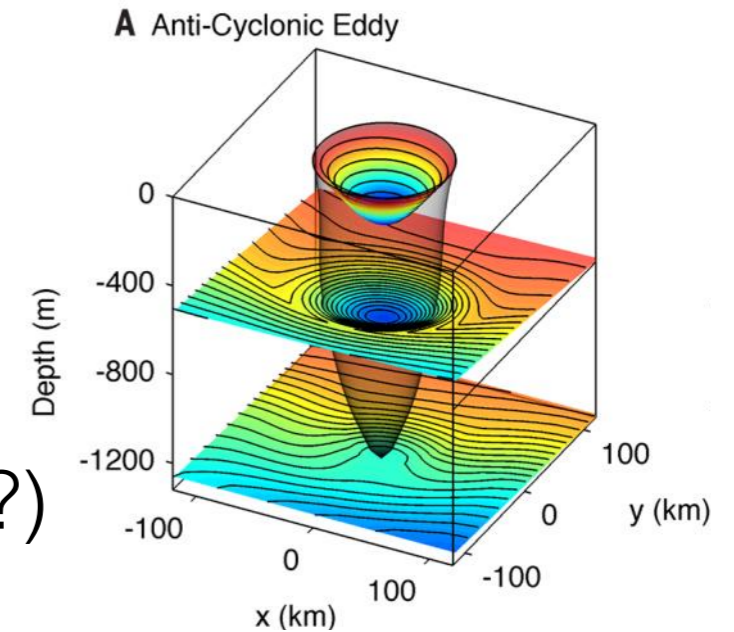
Reynolds Flux

$$\overline{v'\theta'} \quad \overline{u'v'}$$

- Statistical
- Eulerian
- Flux due to correlations btw. fluctuating velocity and tracer
- Atmospheric science, linear theory, eddy parameterization, model diagnostics

Coherent Eddy

- Countable
- Lagrangian(?)
- Flux due to motion of coherent structure
- Dynamical systems, observational physical oceanography



Eddy fluxes from coherent structures

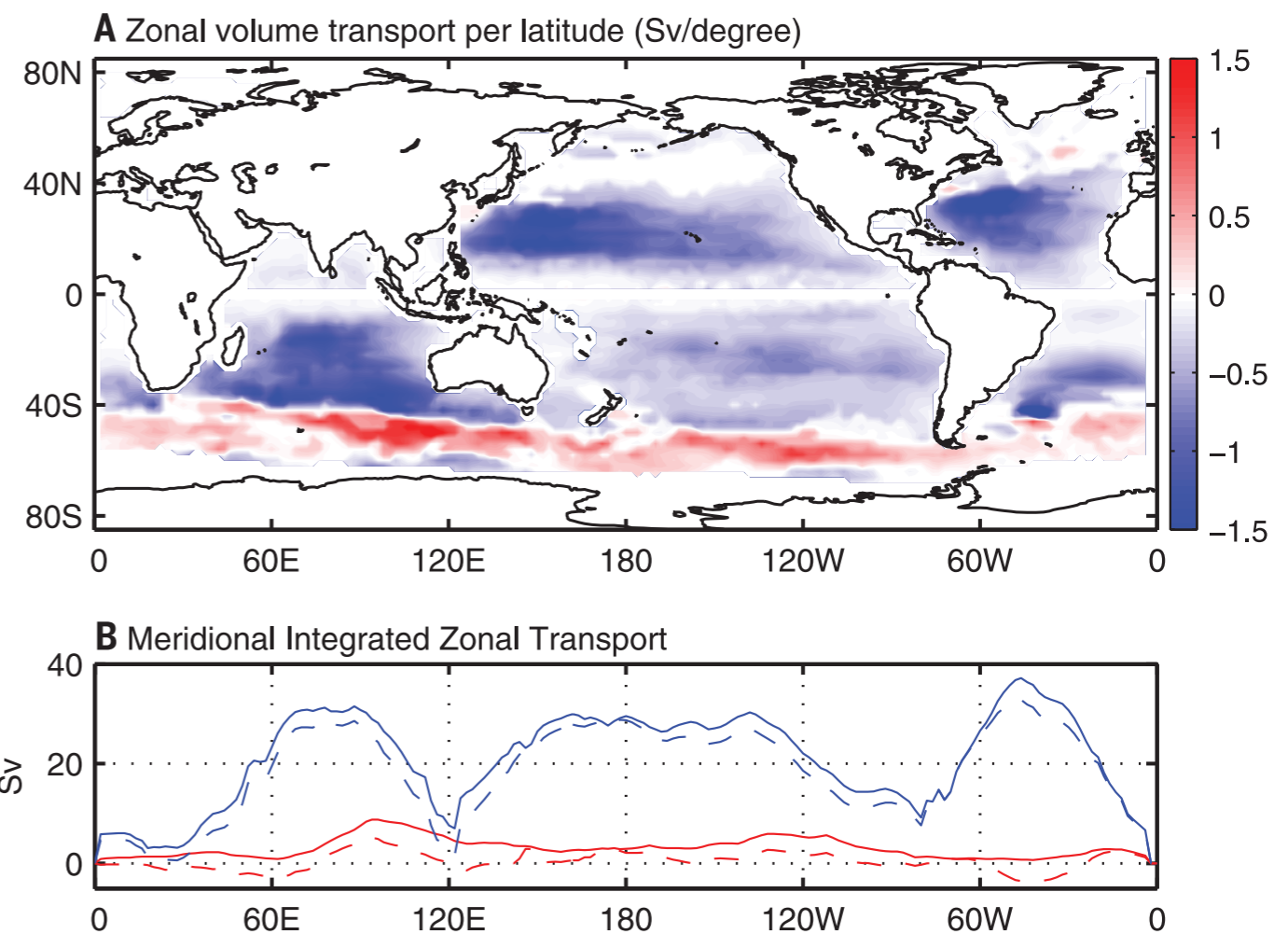
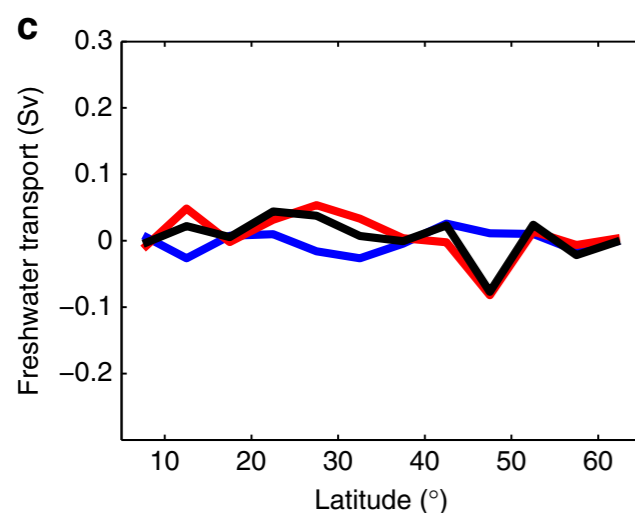
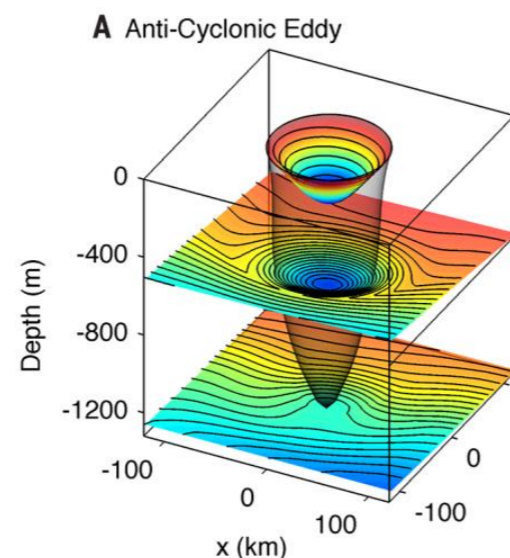
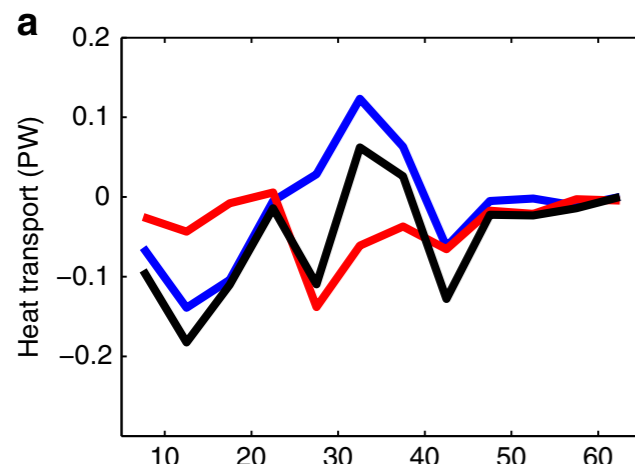
ARTICLE

Received 23 Jun 2013 | Accepted 22 Jan 2014 | Published 18 Feb 2014

DOI: 10.1038/ncomms4294

Global heat and salt transports by eddy movement

Changming Dong^{1,2,3}, James C. McWilliams², Yu Liu³ & Dake Chen¹



“...eddy heat and salt transports are mainly due to individual eddy movements.”

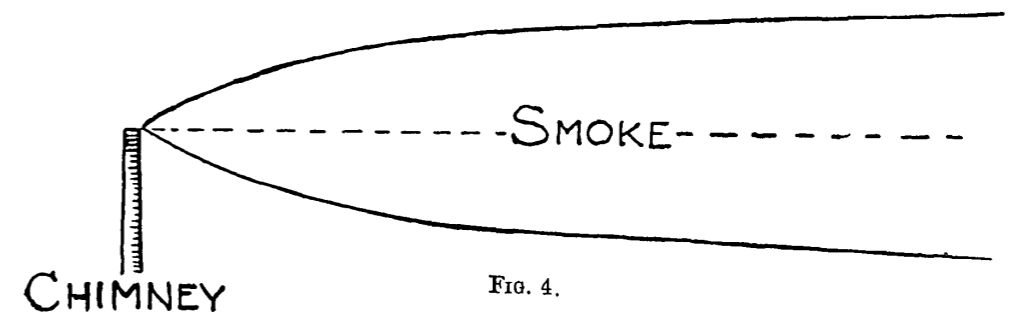
Are all the “eddies” seen in altimetry actually materially coherent?

Lagrangian Dispersion

DIFFUSION BY CONTINUOUS MOVEMENTS

By G. I. TAYLOR.

[Received May 22nd, 1920.—Read June 10th, 1920.]



meridional heat flux $\overline{v'\theta'} = -K \frac{\partial \bar{\theta}}{\partial y}$ $K = \frac{1}{2} \frac{\partial}{\partial t} \overline{(\Delta Y)^2}$ eddy diffusivity

$\Delta Y = Y(x, y, t) - Y(x, y, 0)$ single-particle Lagrangian dispersion

Relative dispersion

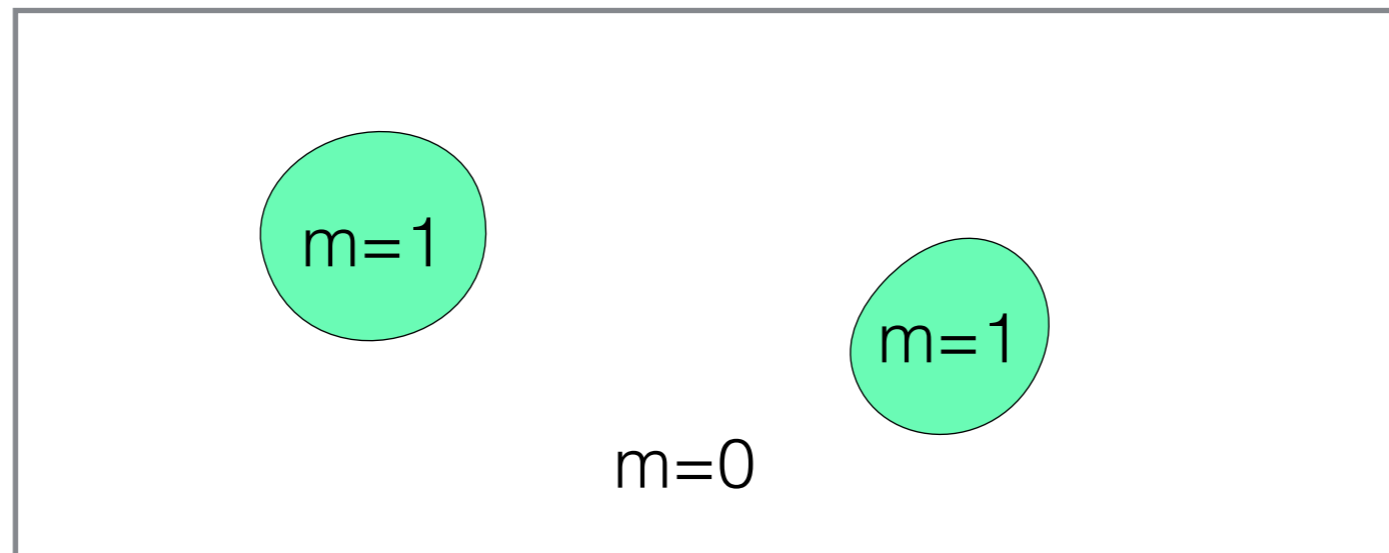
$$K_{rel} = \frac{1}{2} \frac{\partial}{\partial t} \overline{(Y - \bar{Y})^2}$$

- Removes the dispersion due to “mean flow”
- Describes spreading of tracer patch
- Equivalent to K at large separation scales (i.e. long time scales)

Coherent Lagrangian Dispersion

Relative dispersion $K_{rel} = \frac{1}{2} \frac{\partial}{\partial t} \overline{(Y - \bar{Y})^2}$

coherent structure mask



“Coherent” diffusivity $K_{rel}^{CS} = \frac{1}{2} \frac{\partial}{\partial t} \overline{(mY - \bar{Y})^2}$

Assumes that the water outside the CSs is just moving with the mean flow

Coherent Lagrangian Dispersion



256,000 particles

Coherent Lagrangian Dispersion

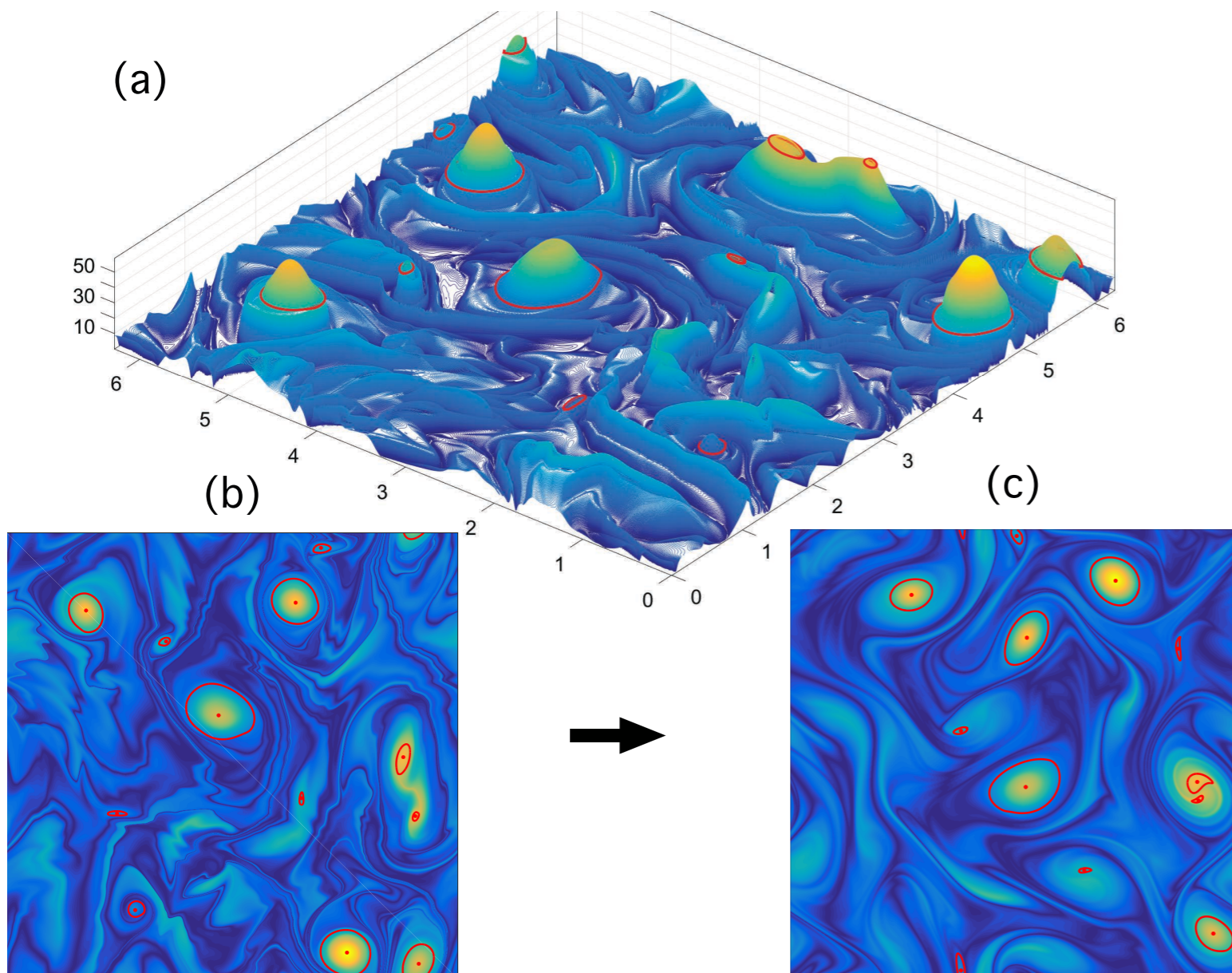


Rotationally Coherent Lagrangian Vortices

G. Haller, A. Hadjighasem, M. Farazamand & F. Huhn, 2016
Defining coherent vortices objectively from the vorticity. J. Fluid Mech.

Lagrangian-Averaged Vorticity Deviation:

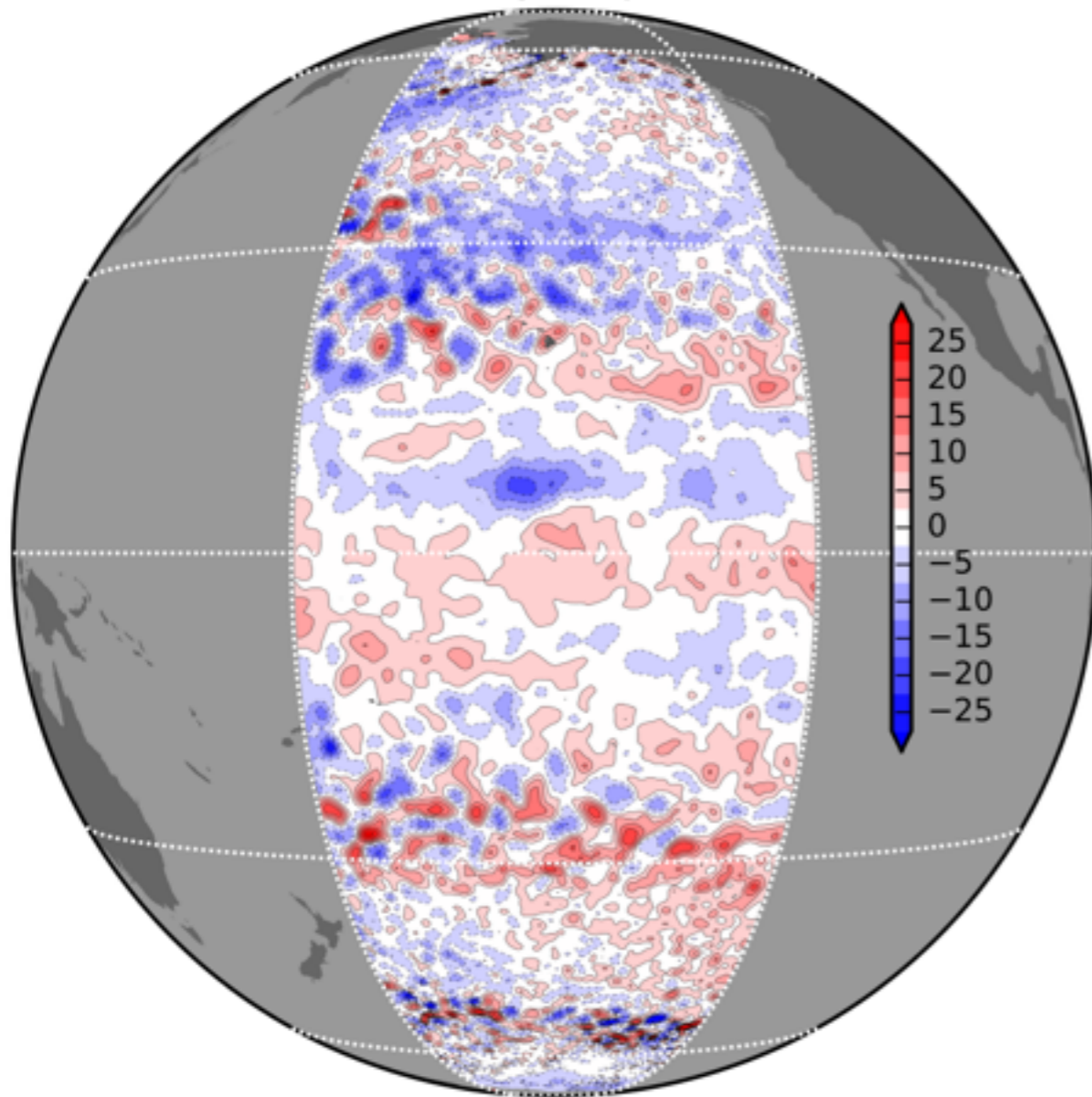
$$\text{LAVD}_{t_0}^t(x_0) := \int_{t_0}^t |\omega(x(s; x_0), s) - \bar{\omega}(s)| ds \quad \text{integral taken along Lagrangian trajectory}$$



- Vortex centers: maxima of $|\text{LAVD}|$
- Vortex boundaries: outermost *convex* curves around maxima
- Coherence is defined over a *finite time interval*

East Pacific Sector

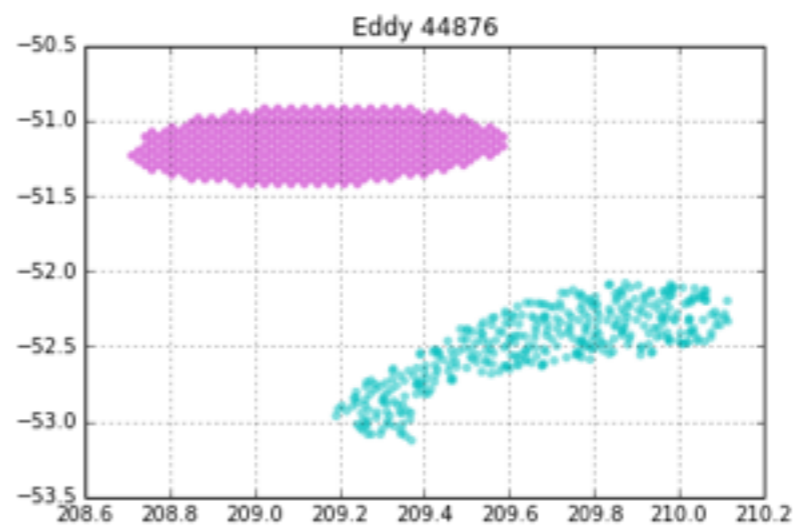
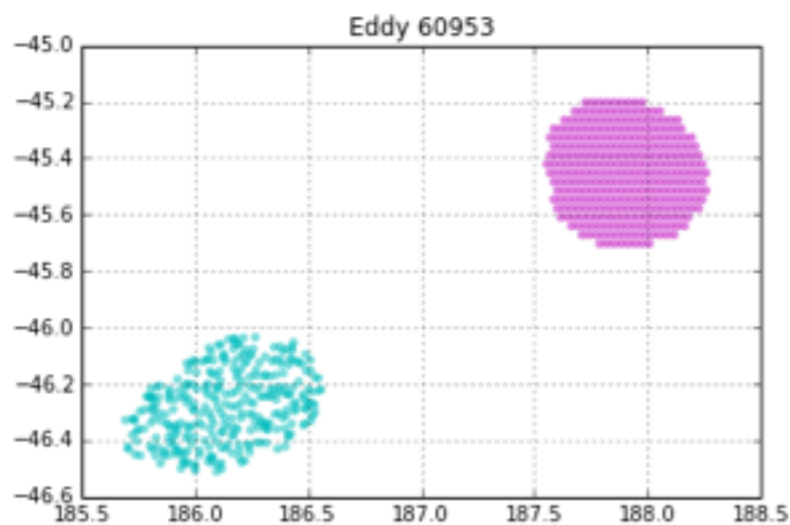
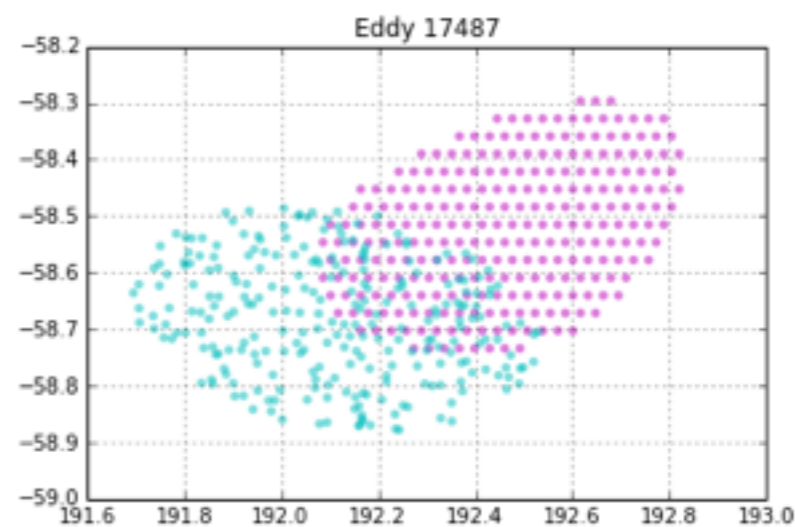
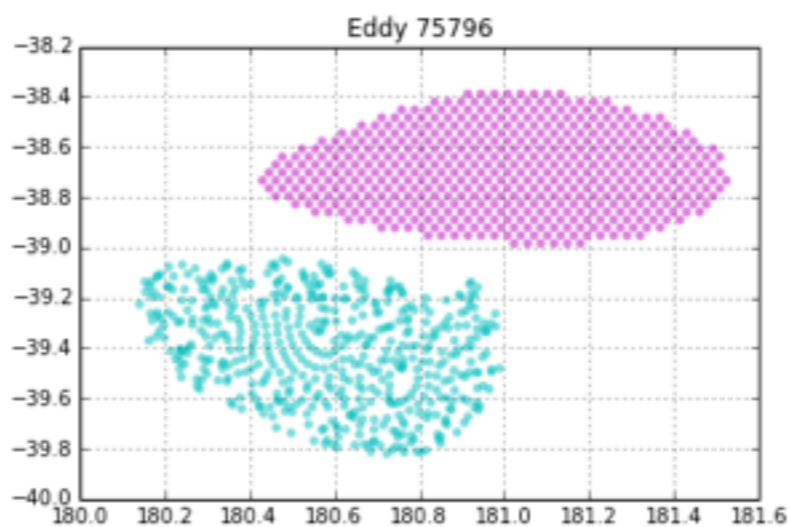
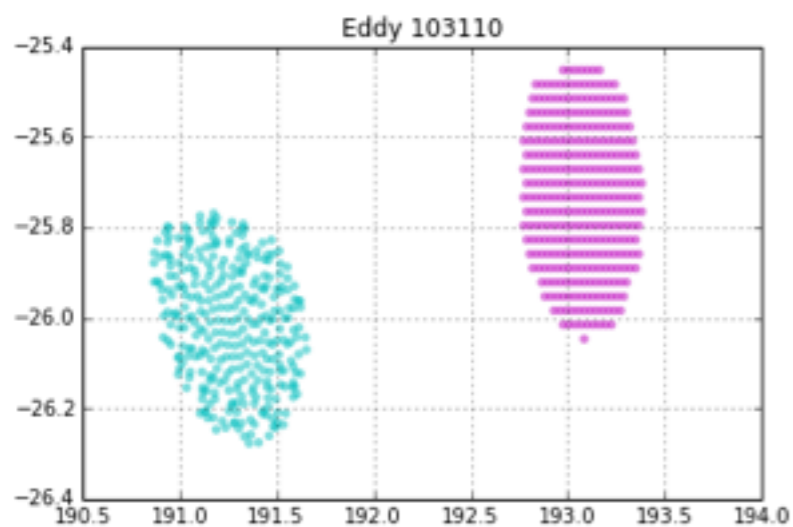
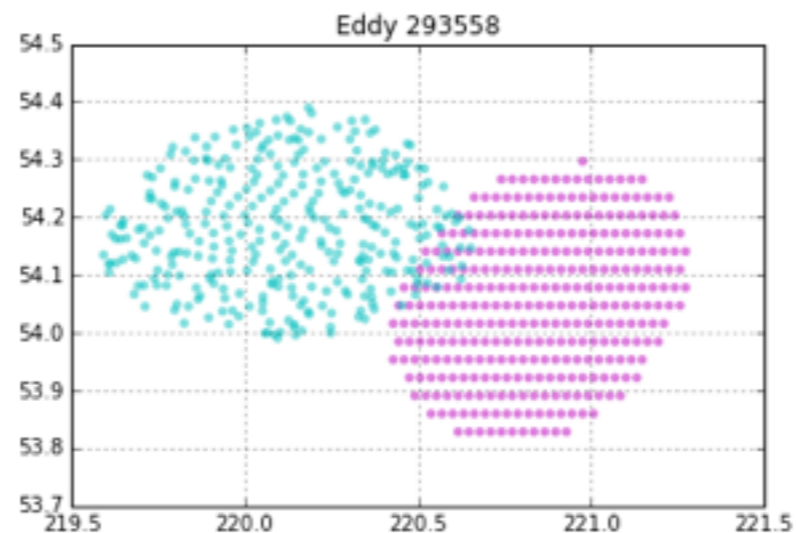
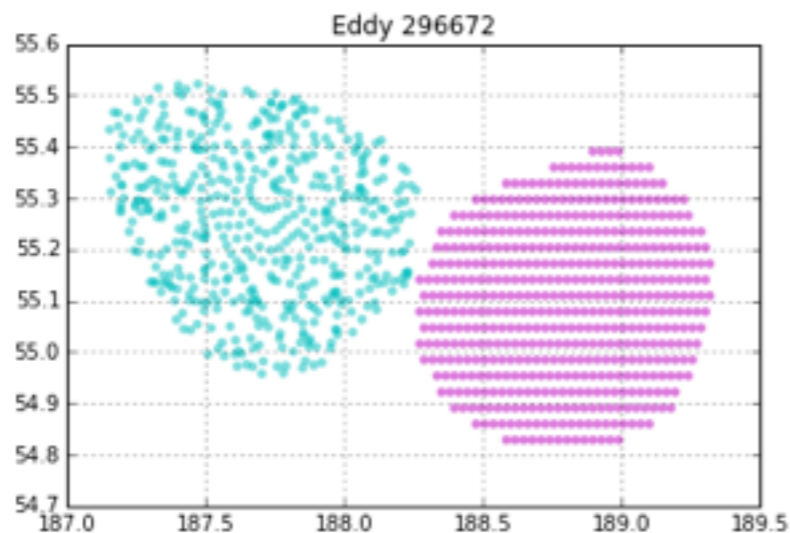
SSH Anomaly Snapshot (cm)



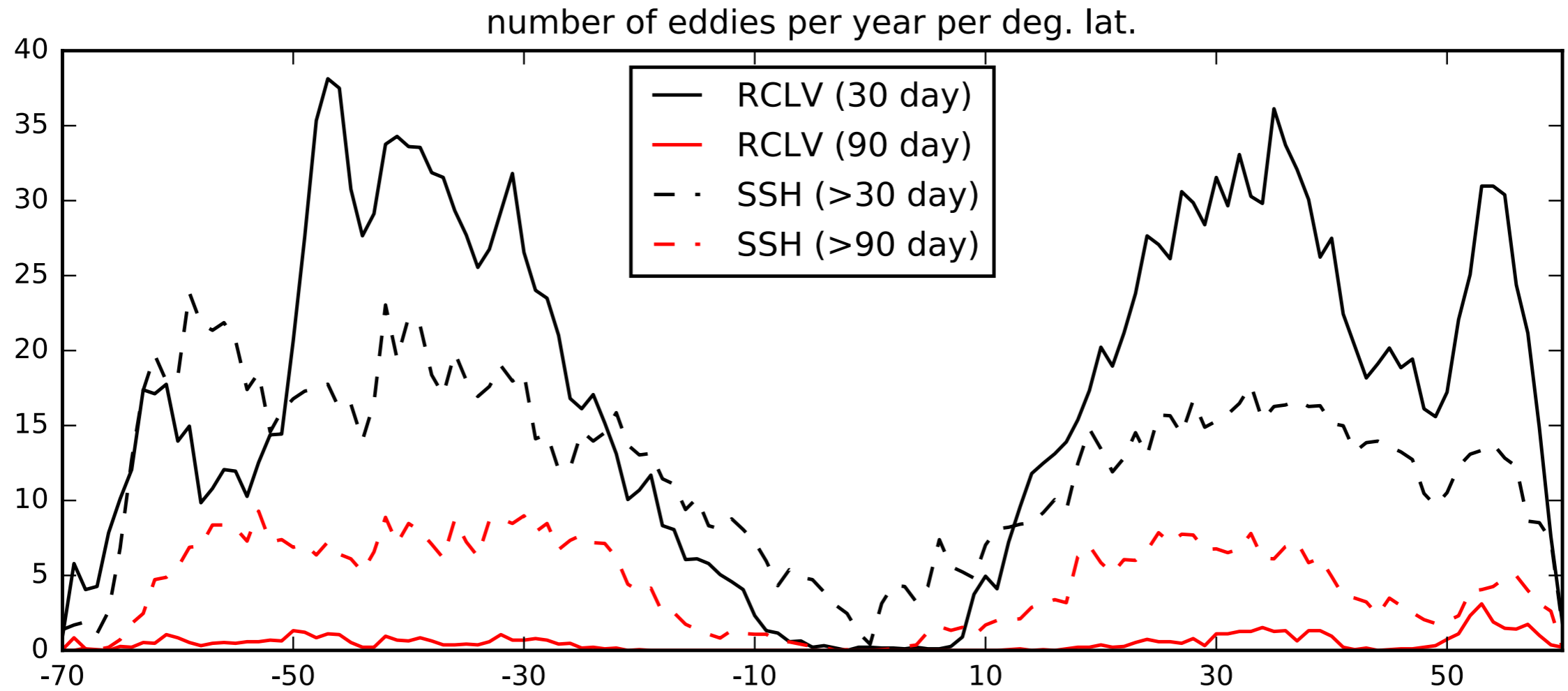
$$\hat{\mathbf{k}} \times \mathbf{v}_g = -\frac{g}{f} \nabla \eta$$

- Very little land, mostly zonally symmetric statistics
- Eulerian eddy fluxes are very well studied (Abernathy & Marshall, 2013; Klocker & Abernathy, 2014; Abernathy & Wortham, 2015)
- Use AVISO altimetric velocities to advect Lagrangian particles (30, 90, 270 days)
- 1/32-degree mesh is necessary to resolve RCLVs: ~8 million particles
- Track position, relative vorticity, of particles with daily sampling

90-day RLCVs

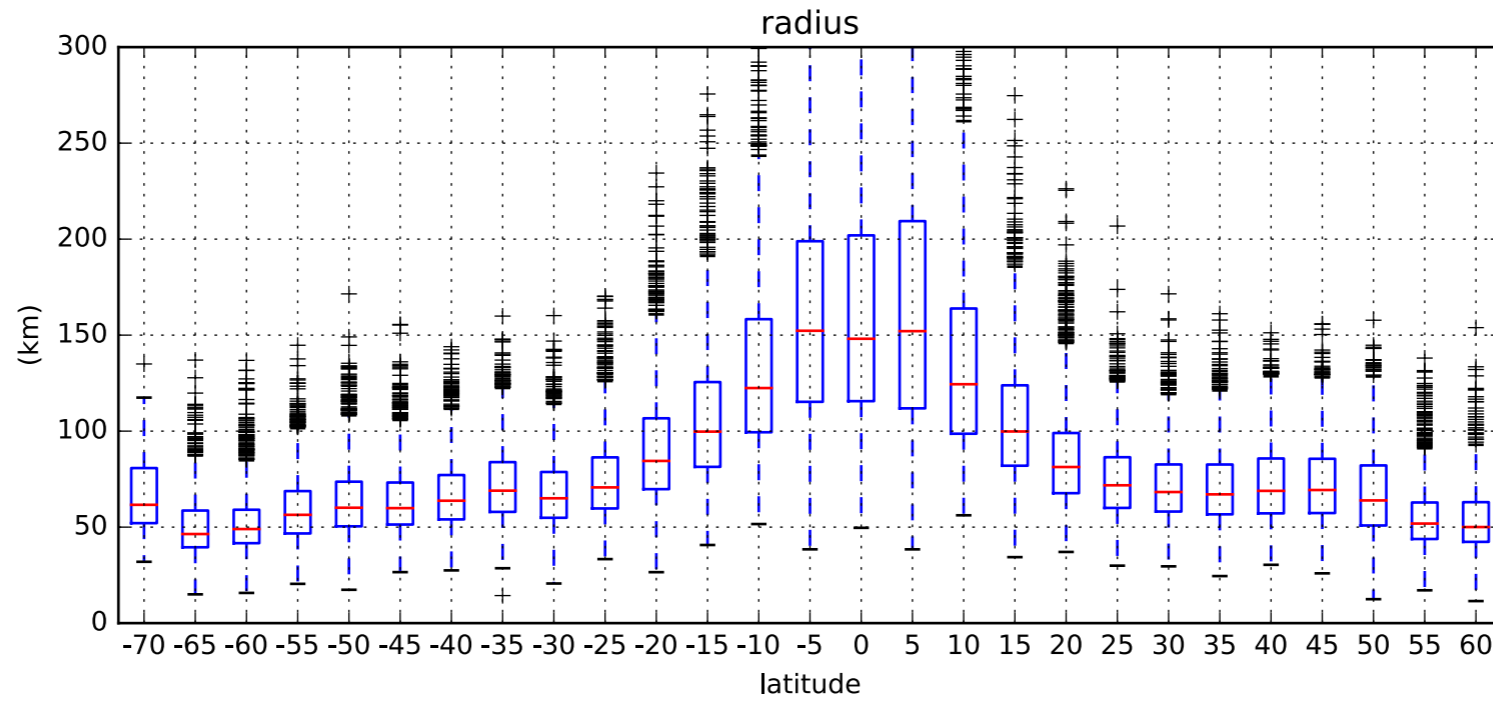


Eddy Statistics: Frequency

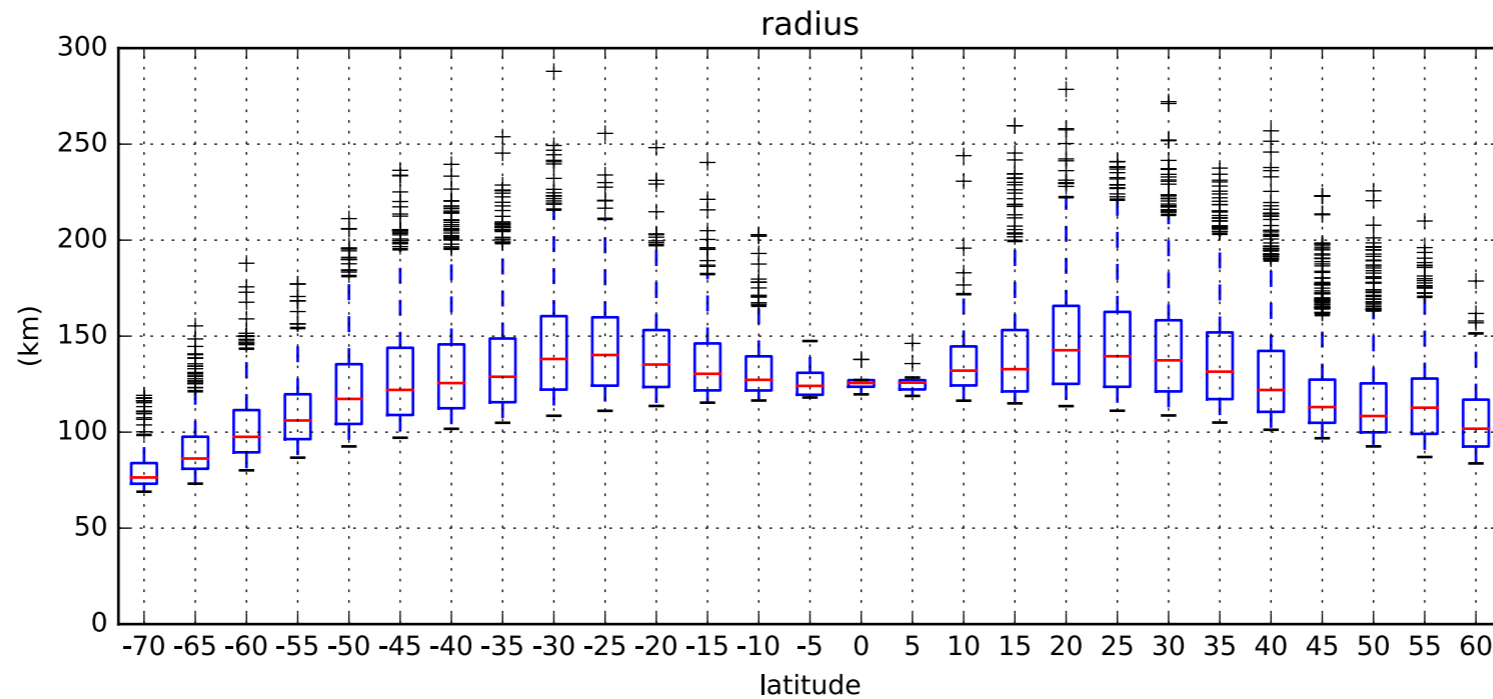


Eddy Statistics: Radius

SSH



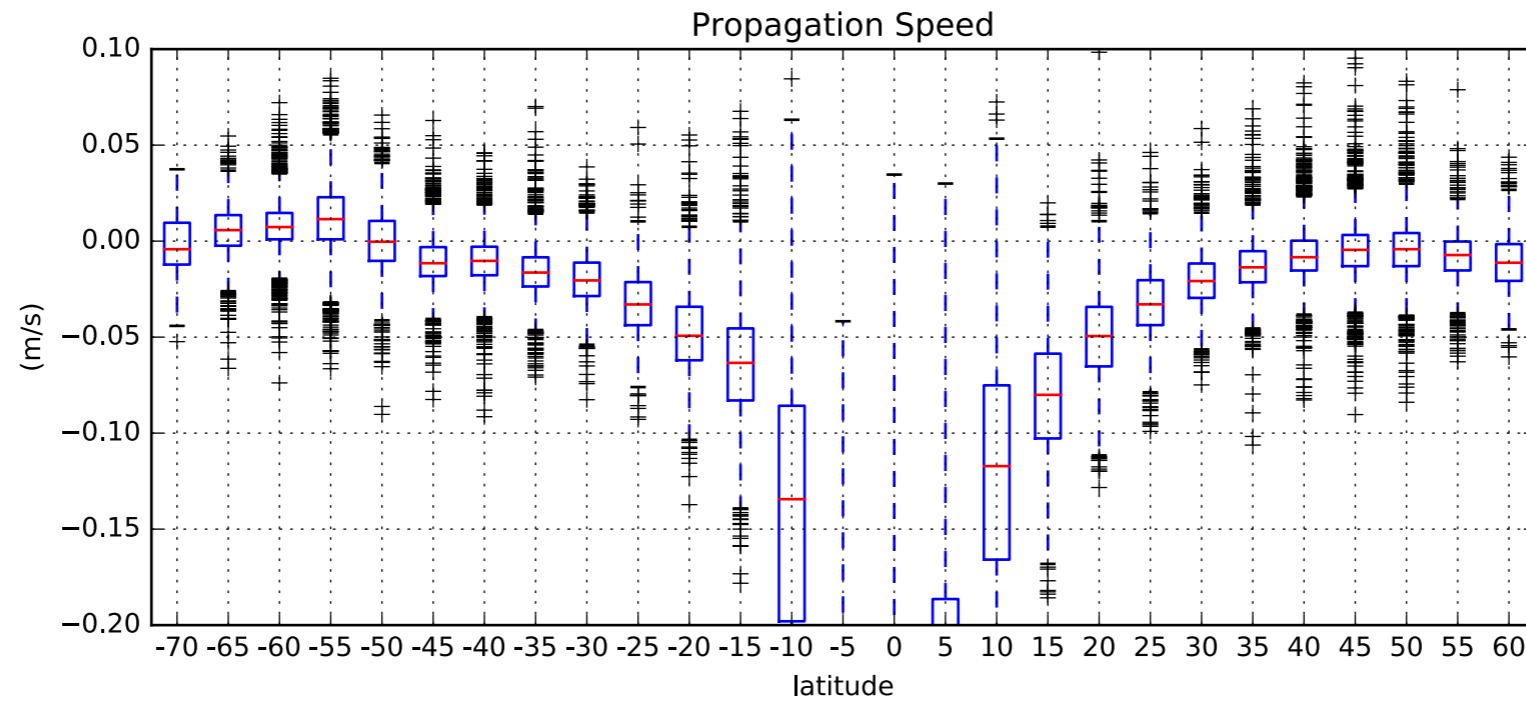
RCS
(30 day)



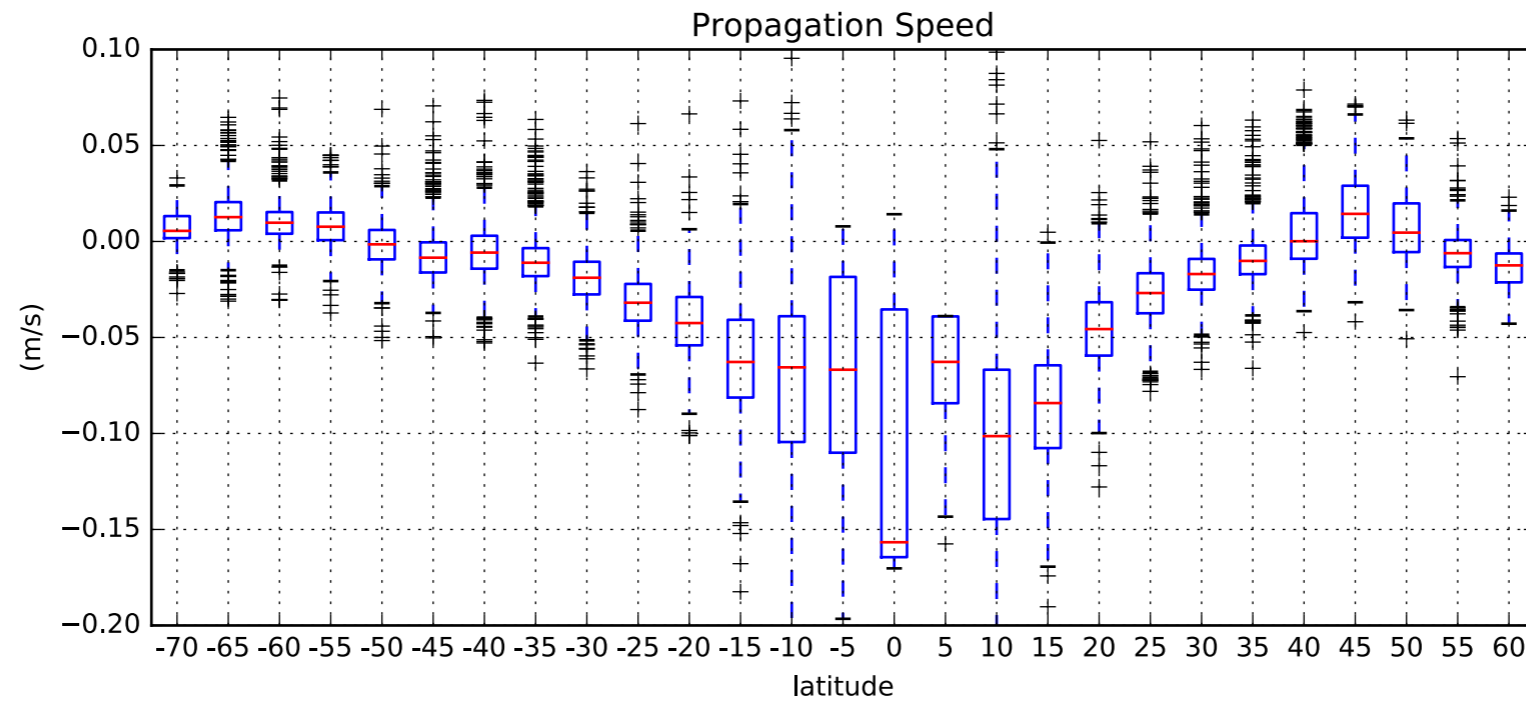
“mesoscale”

Eddy Statistics: Zonal Propagation

SSH

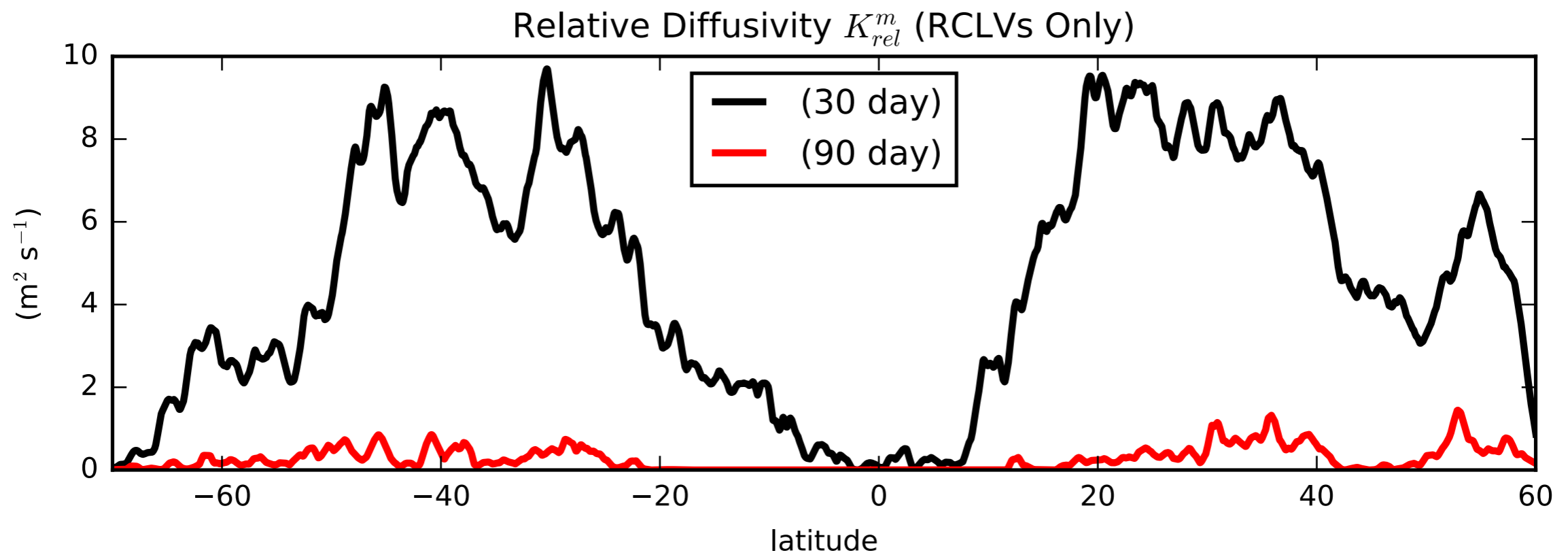
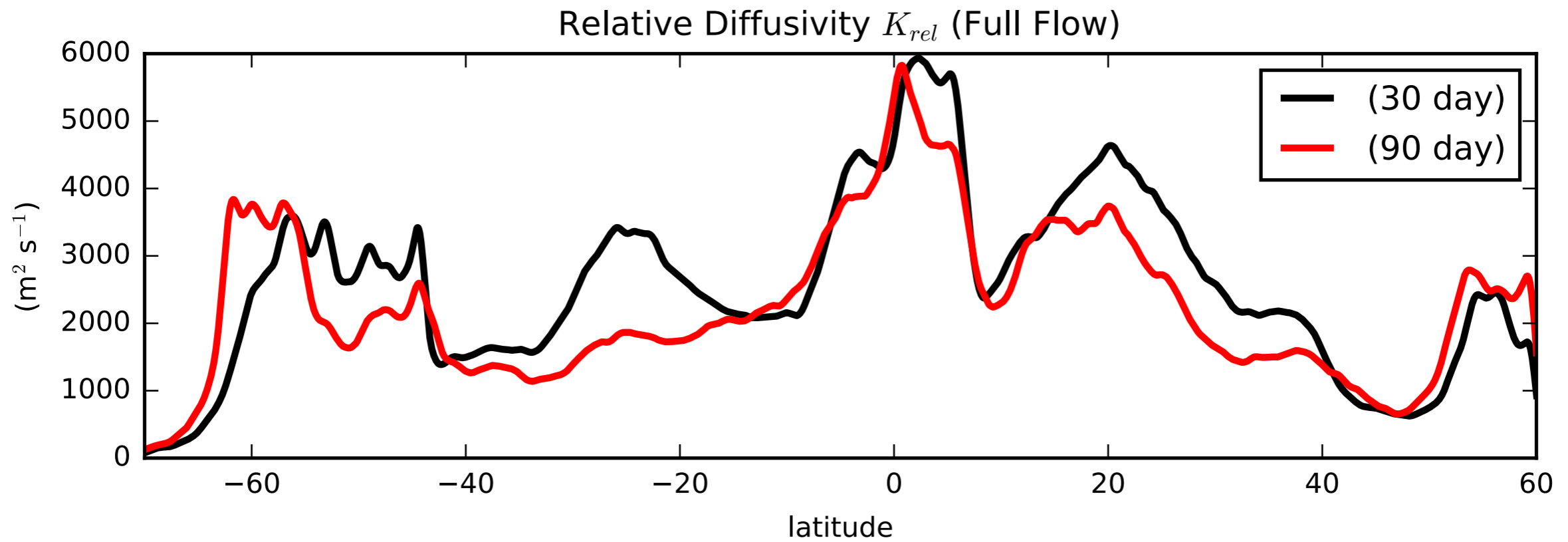


RCS
(30 day)

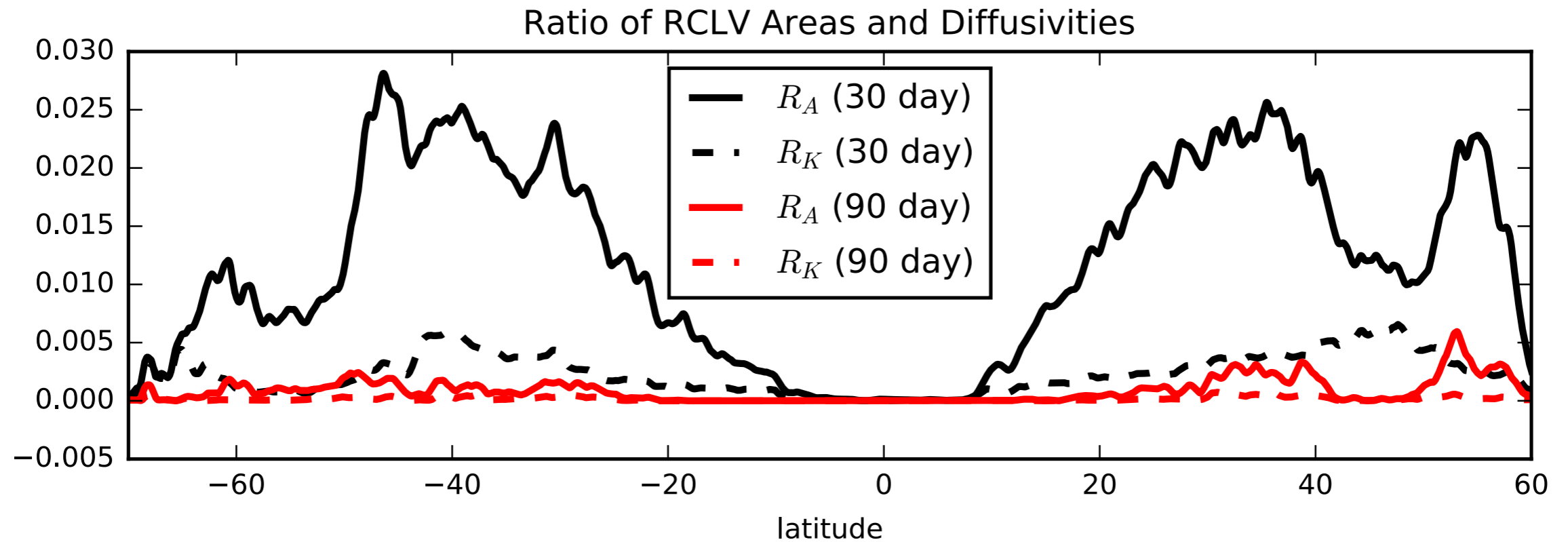


both “feel”
Rossby wave
dispersion
relation

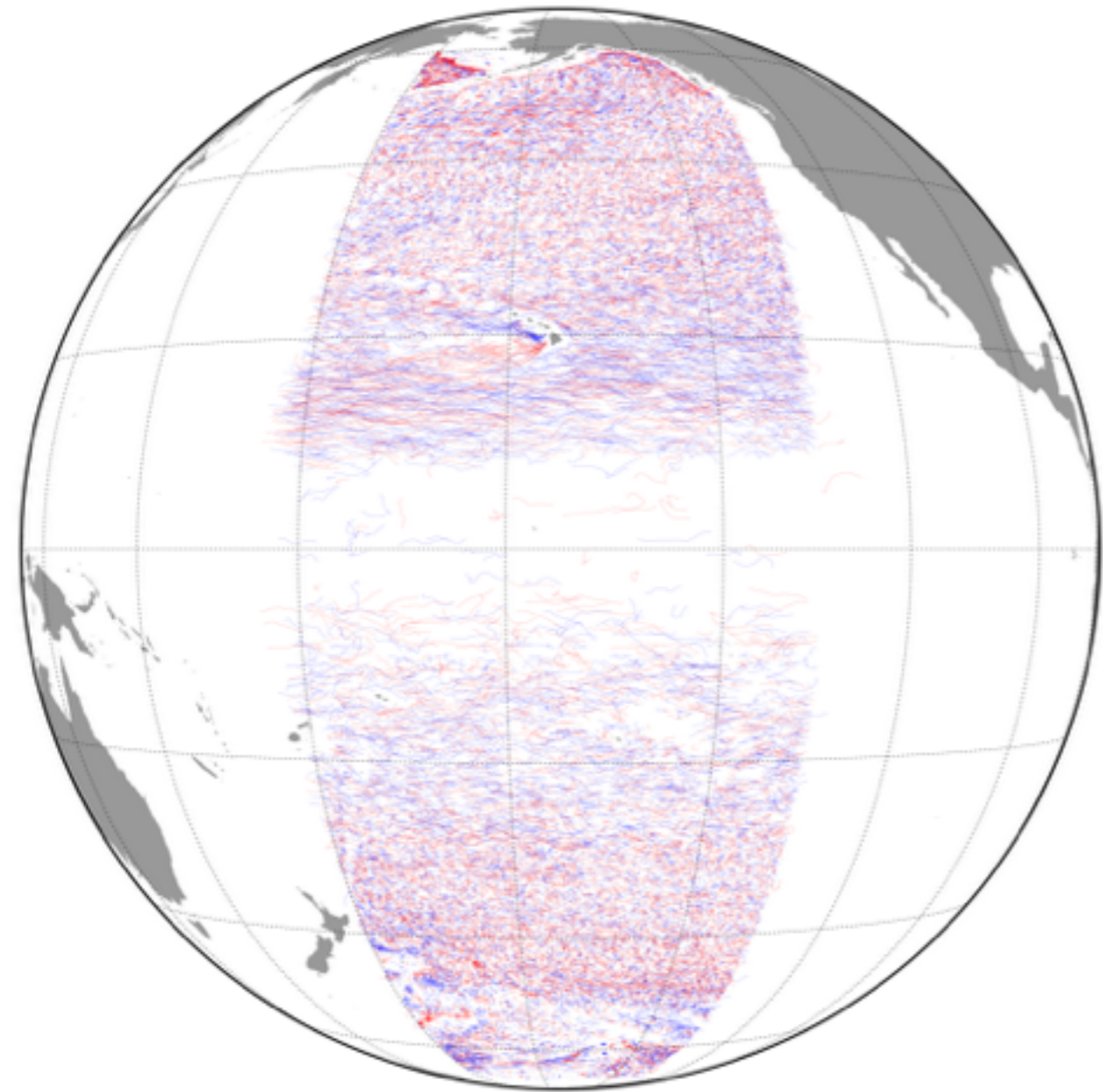
Diffusivity



Area and Diffusivity Ratios



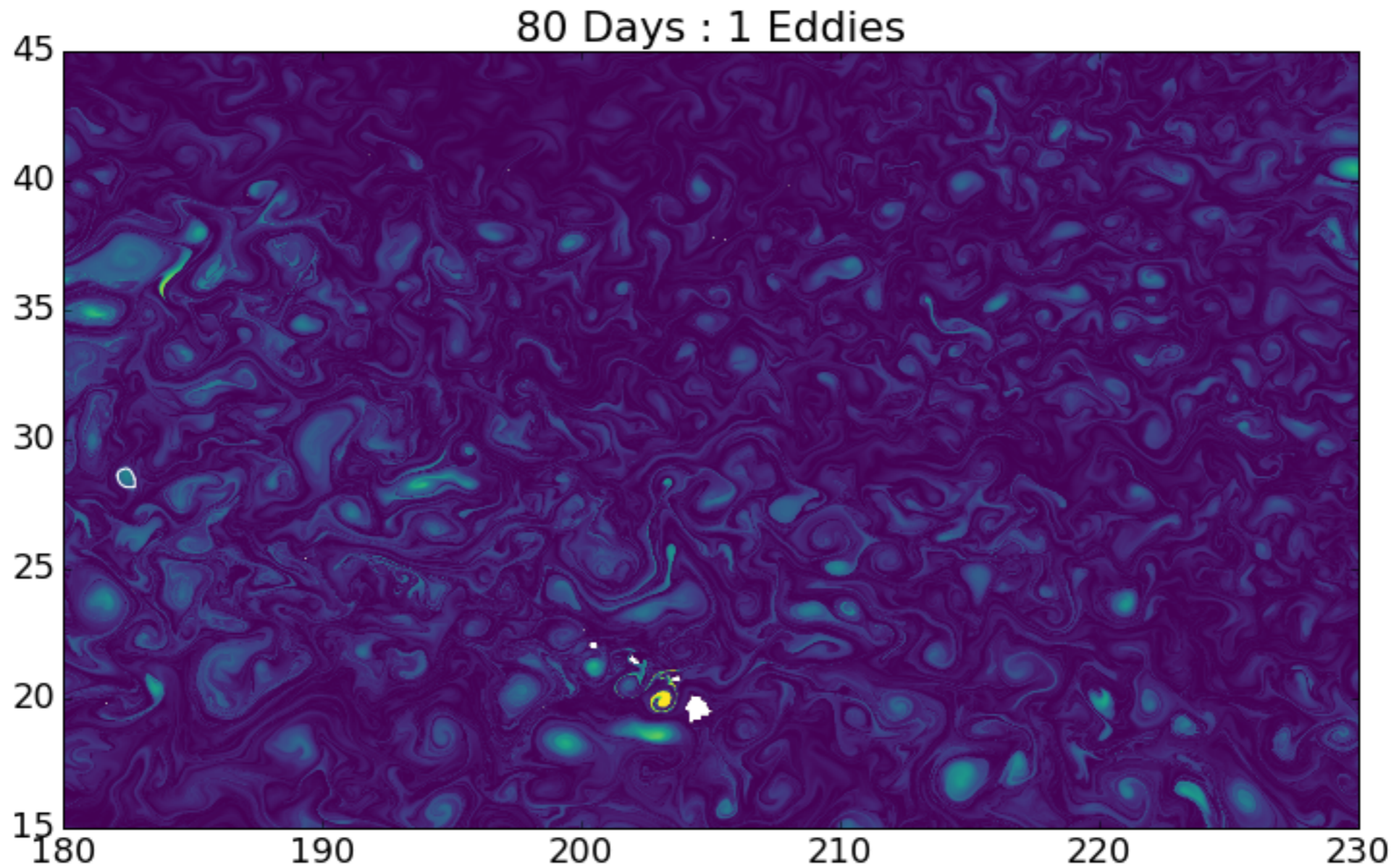
Conclusions



- It is feasible to perform a comprehensive “census” of RCLVs on a global scale using the entire altimetric dataset
- Certain statistics of RCLVs are broadly similar to SSH-eddies (e.g. propagation speed)...
- ...But the RCLVs are much shorter lived
- Calculated the contributions RCLVs to net meridional dispersion (relative diffusivity): **totally negligible!**
- More validation and analysis in progress

Identifying RLCVs

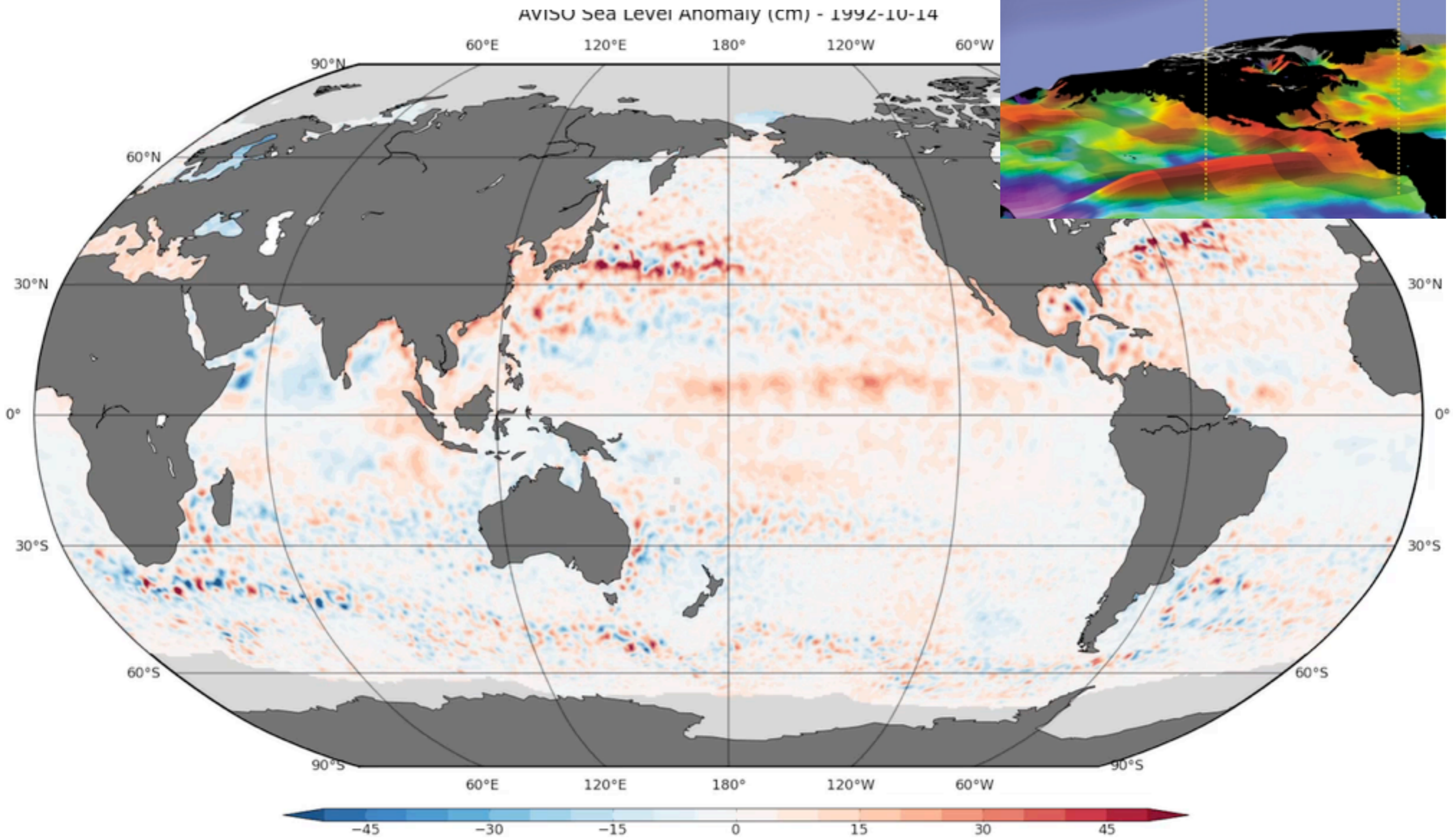
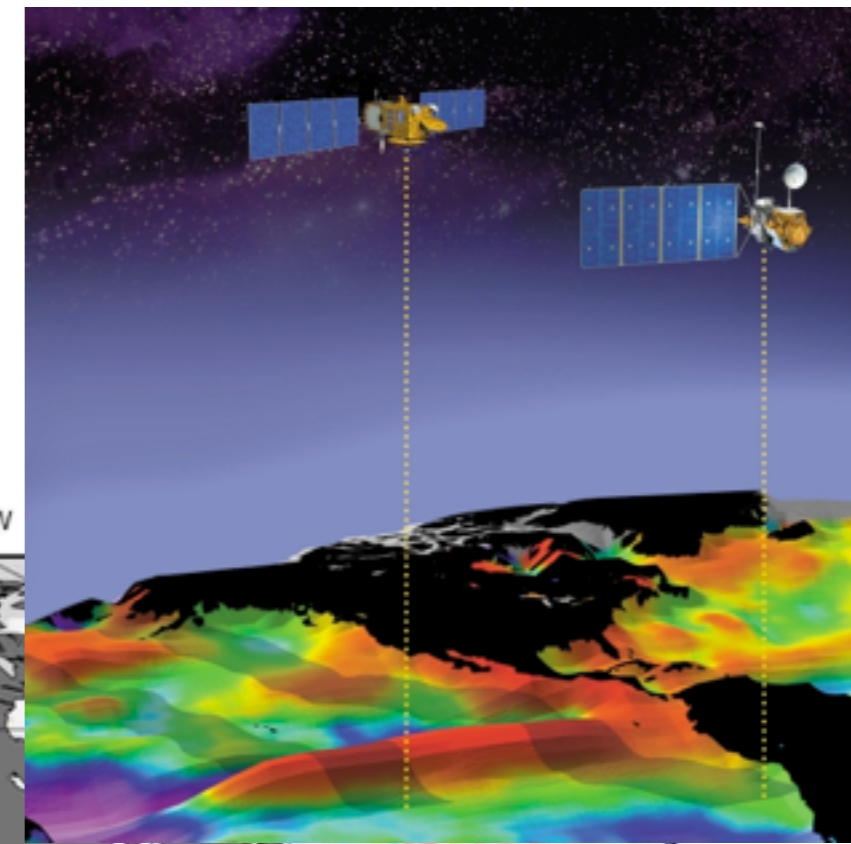
dependence on time interval



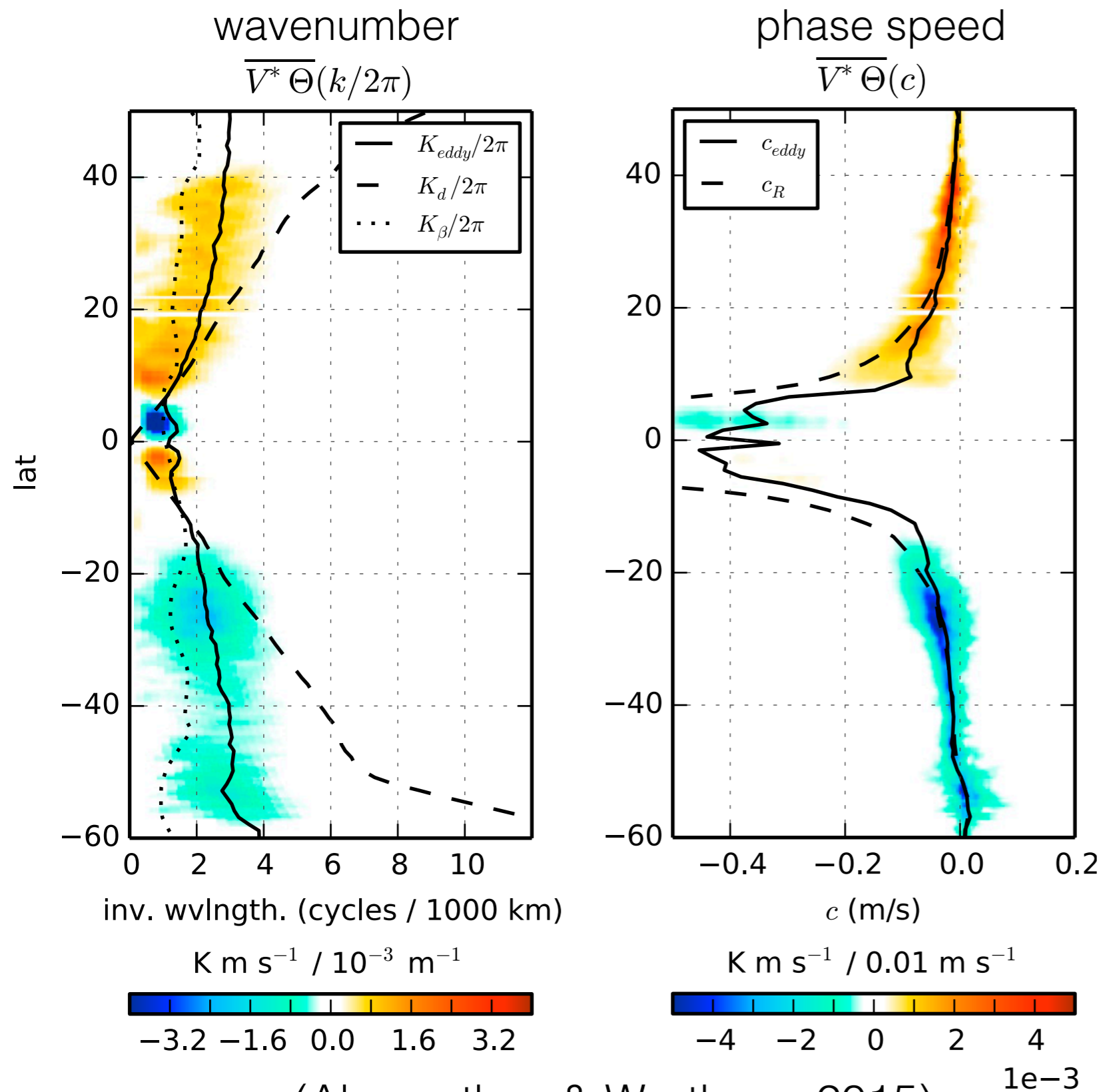
satellite altimetry

SSH is streamfunction
for geostrophic flow

$$\hat{\mathbf{k}} \times \mathbf{v}_g = -\frac{g}{f} \nabla \eta$$



East Pacific Sector - Heat Flux from Satellite Obs.



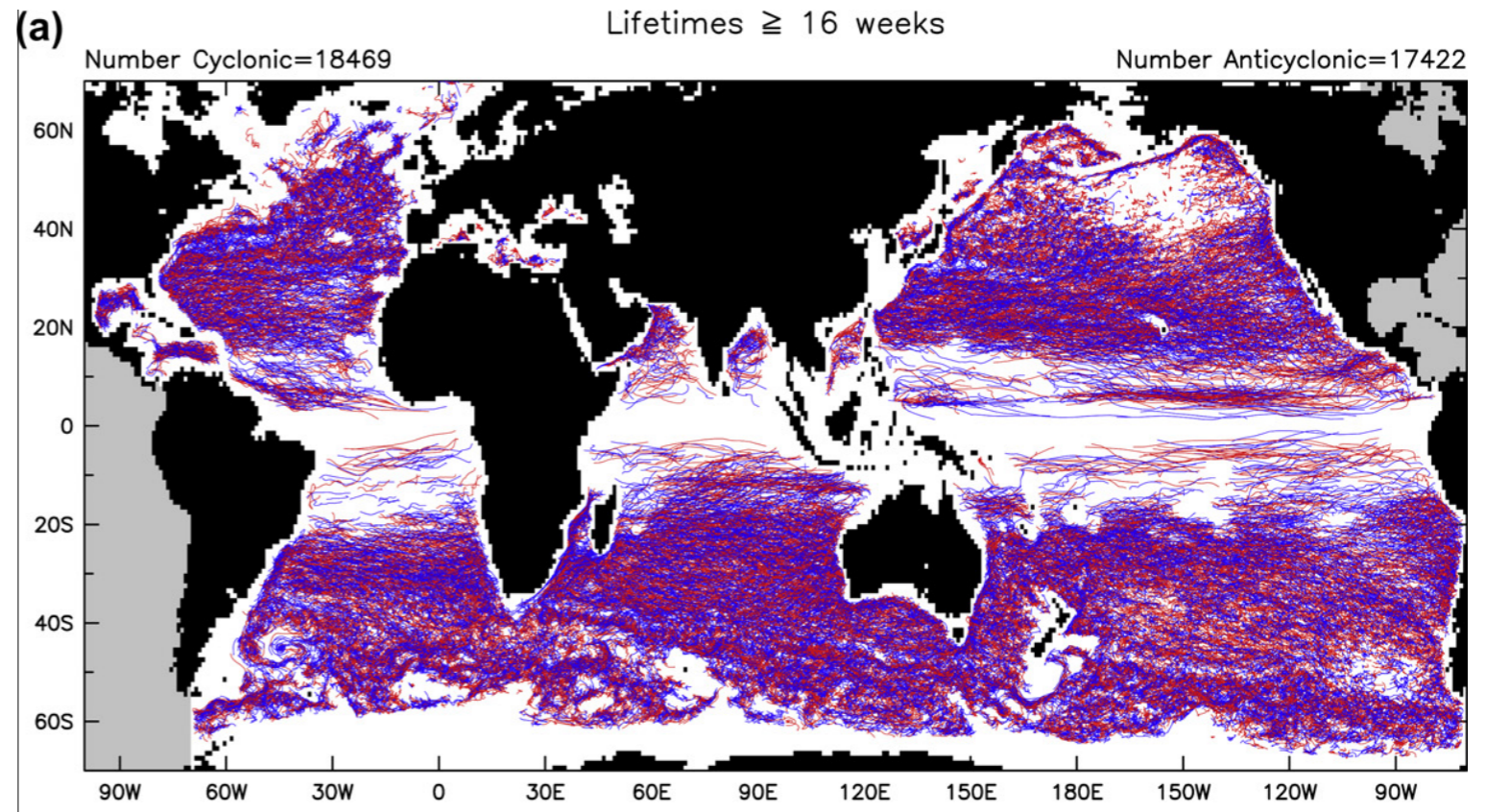
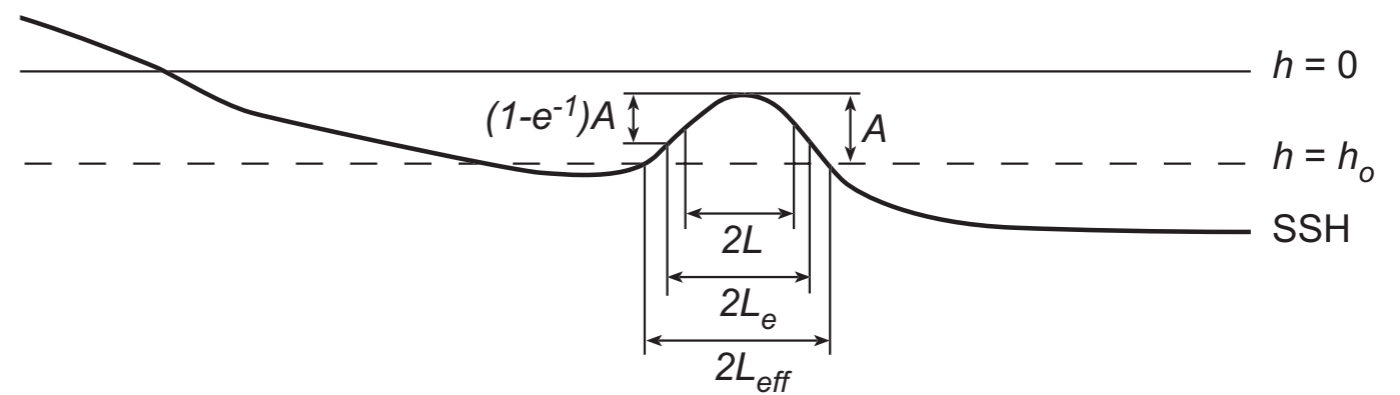
- Cross-spectral analysis of SST and SSH-derived velocity
- Spectral peaks are aligned with the properties of tracked SSH eddies (Chelton)
- Implies an important role for coherent SSH “features” in eddy heat transport

(Abernathy & Wortham, 2015)

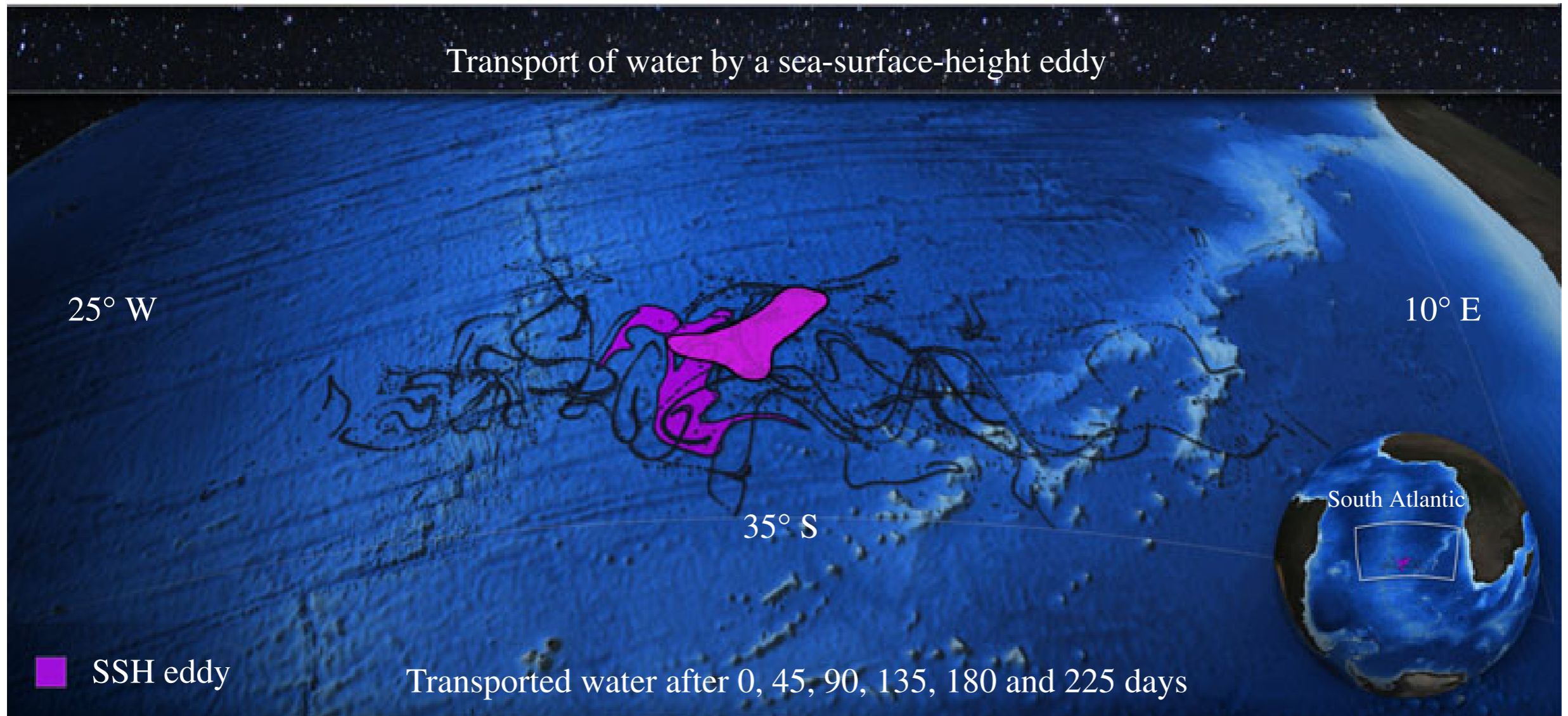
Sea-surface-height eddies

(Chelton et al., 2011)

1. The SSH values of all of the pixels are above (below) a given SSH threshold for anticyclonic (cyclonic) eddies.
2. There are at least 8 pixels and fewer than 1000 pixels comprising the connected region.
3. There is at least one local maximum (minimum) of SSH for anticyclonic (cyclonic) eddies.
4. The amplitude of the eddy is at least 1 cm (see below).
5. The distance between any pair of points within the connected region must be less than a specified maximum.



...But SSH eddies are NOT materially coherent



“Coherently evolving velocity features tend to differ substantially from coherently moving fluid parcels...in unsteady flows”
Haller & Beron Vera (2013)

Mathematicians to the rescue...

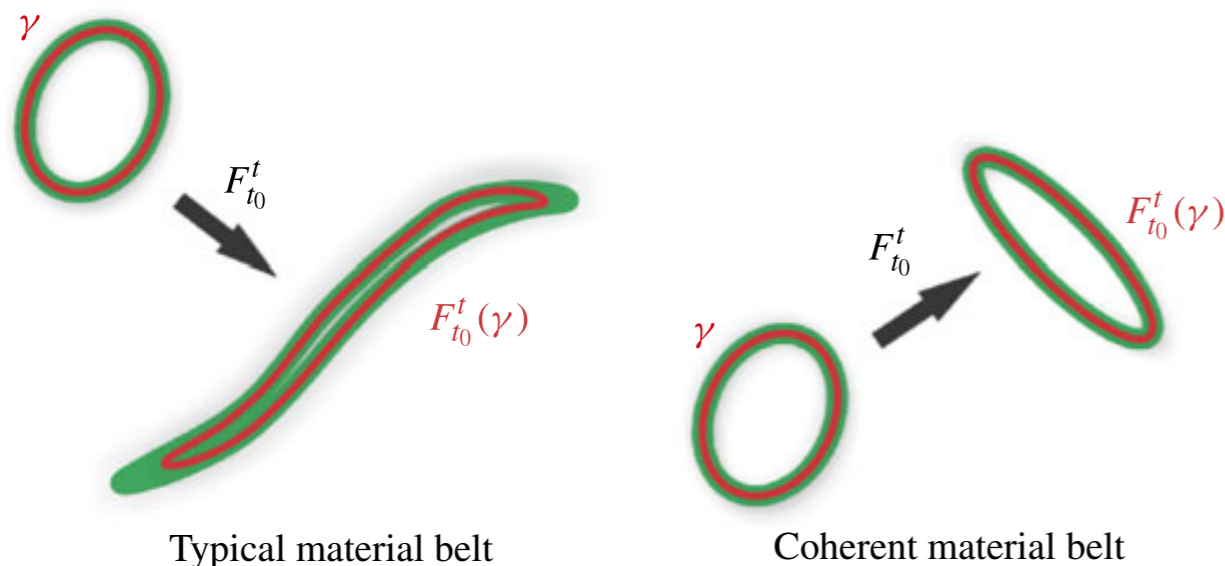


Apply dynamical systems theory to the problem of eddy detection.

Method must be:

- Objective / frame-invariant (observers in all reference frames would find the same eddies)
- Free of arbitrary “tuning parameters”

Solution: use Lagrangian “flow map” to identify exceptionally coherent material lines \Rightarrow *Lagrangian Coherent Structures* (LCS)



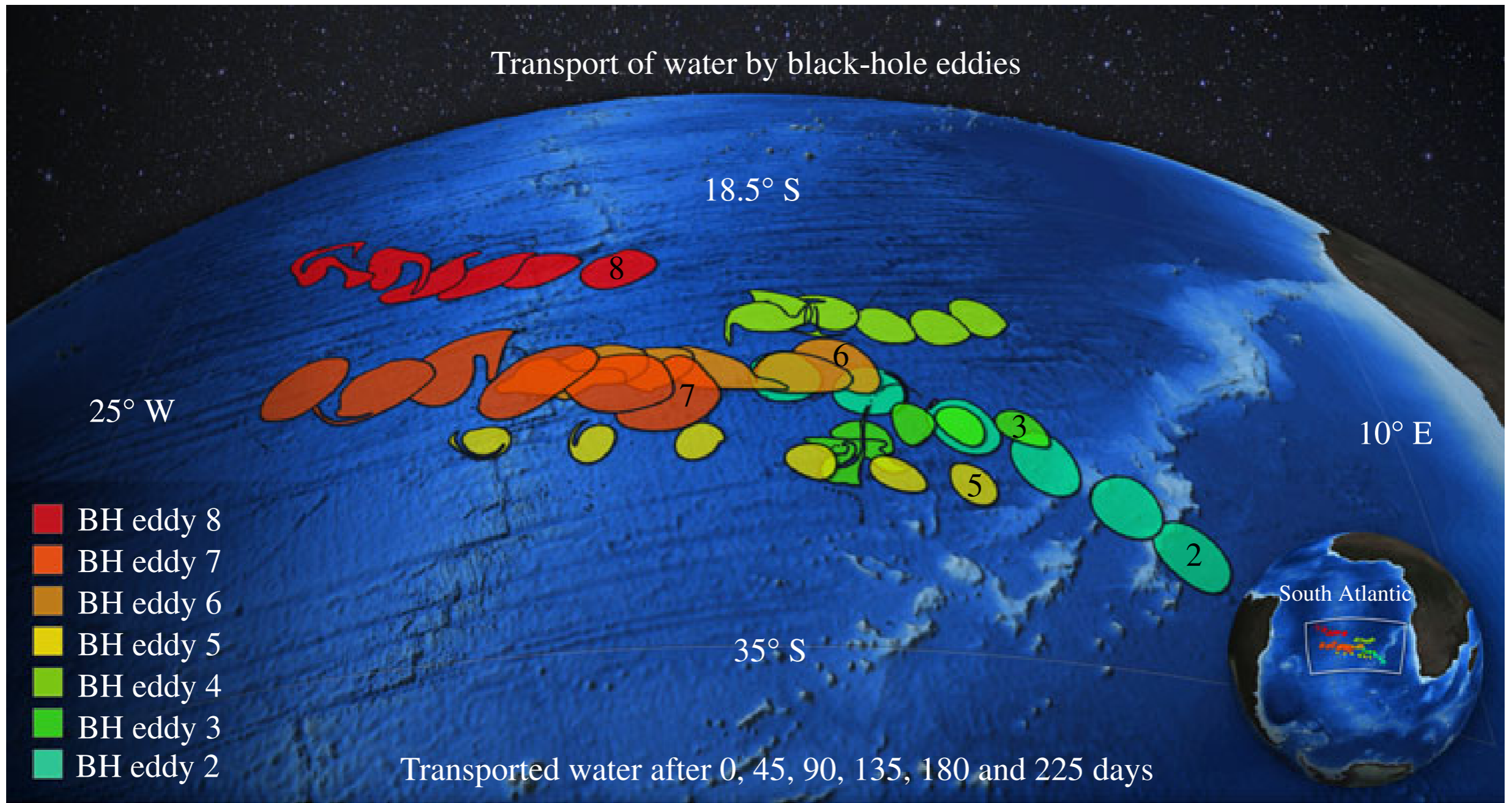
flow map:

$$F_{t_0}^t(x_0) := x(t; t_0, x_0), \quad x \in U, \quad t \in [t_0, t_1],$$

Cauchy-Green strain tensor:

$$C(x_0) = [\nabla F_{t_0}^{t_1}(x_0)]^T \nabla F_{t_0}^{t_1}(x_0).$$

Mathematicians to the rescue...



Computing The Flow Map

initial position $\mathbf{x}_0 \rightarrow \mathbf{x}_1$ *final position*

- Initialize a dense “mesh” of Lagrangian particles over the whole domain (~8 million particles)
- Solve the equation $\frac{d}{dt}\mathbf{x} = \mathbf{v}_g$ for each particle.
- Run forward for 30, 90, 270 days.
- Repeat for each segment of the ~22 year dataset (1993 - 2015)
- ~10 TB of data generated

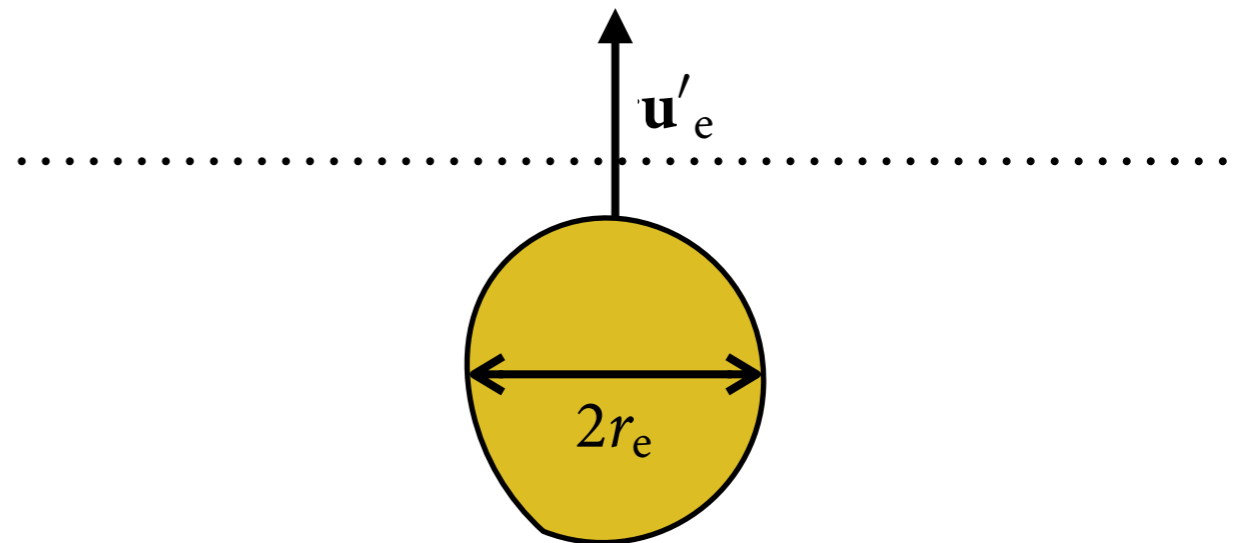
“Transport” by a coherent eddy

$$\mathbf{Q}_{\text{eh}} = s\mathbf{u}'_e \int dz \rho_o C_{\text{po}}(2r_e) T'_e,$$

$$\mathbf{T}_h = N_e \langle \mathbf{Q}_{\text{eh}} \rangle,$$

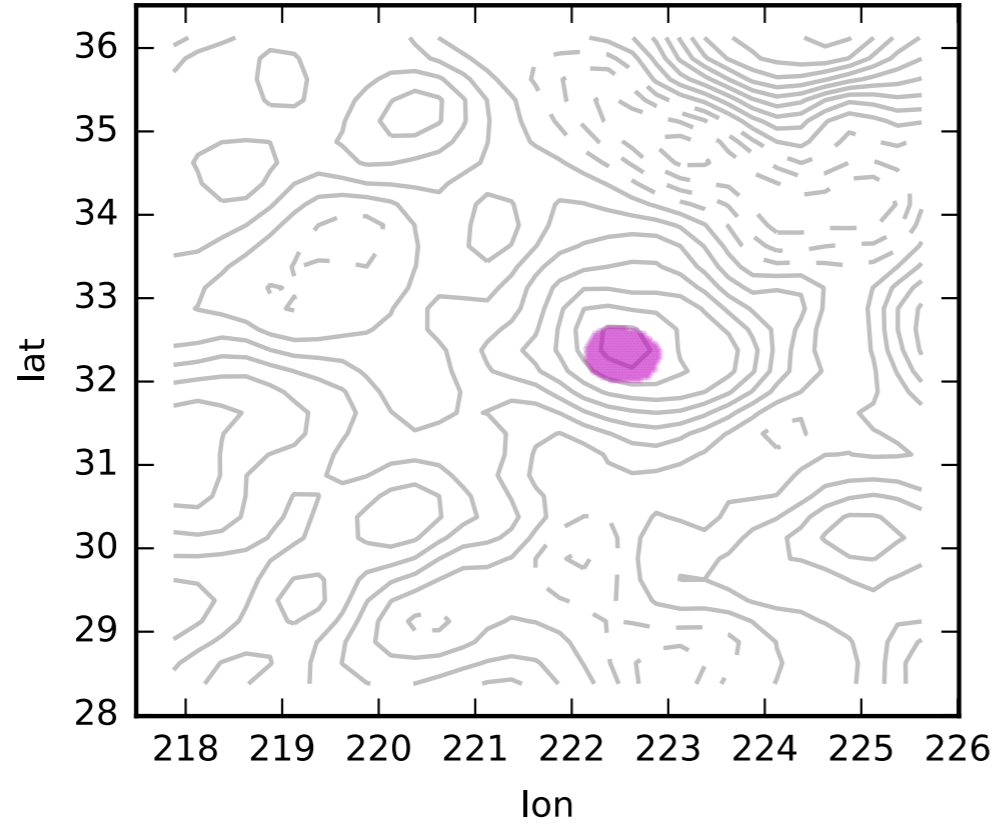
“...eddy heat and salt transports are mainly due to individual eddy movements.”

Dong et al. (2014)



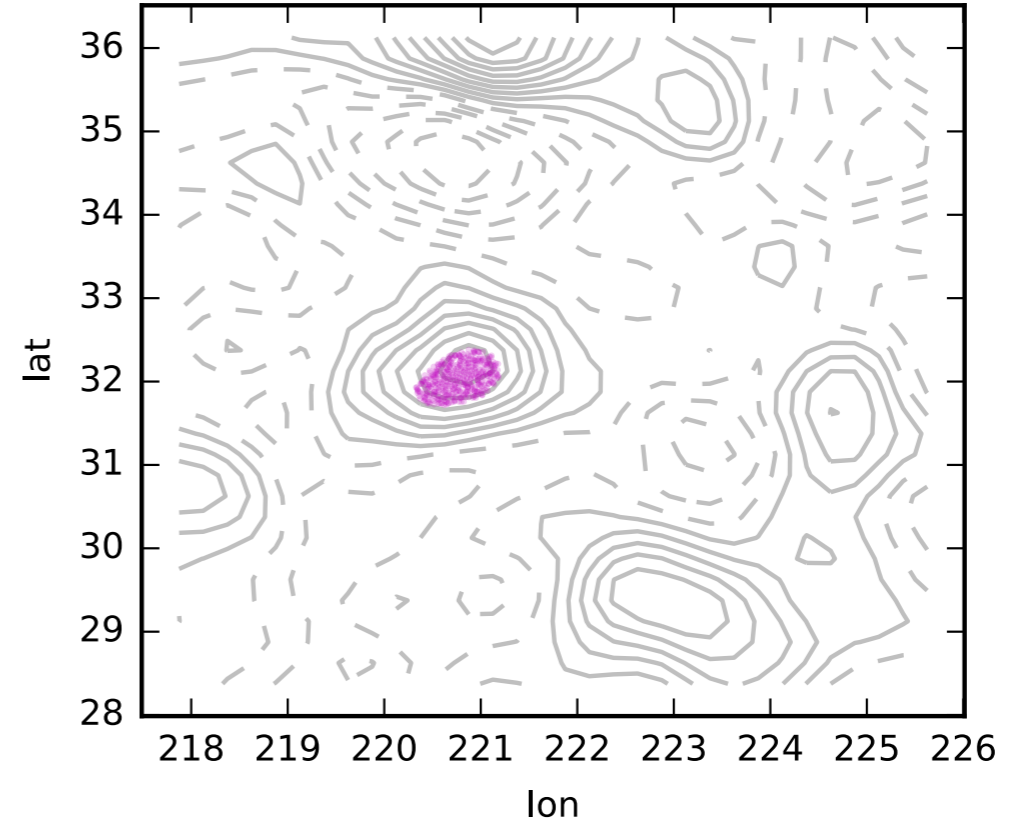
initial

2005-01-28

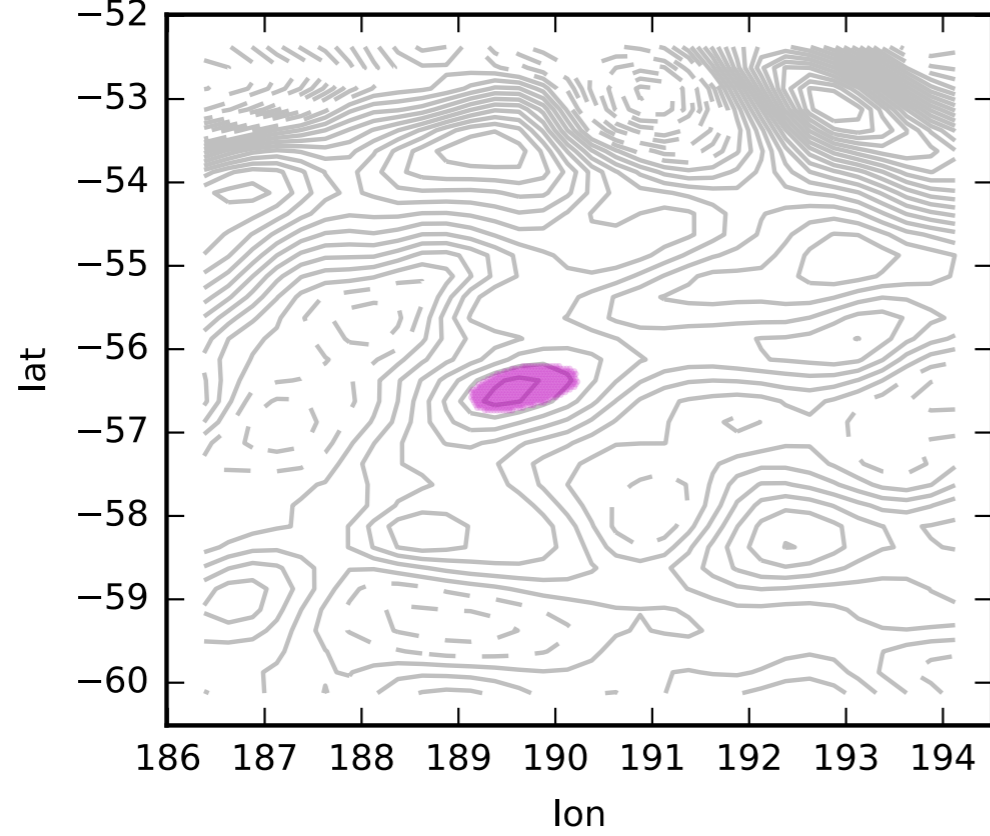


final

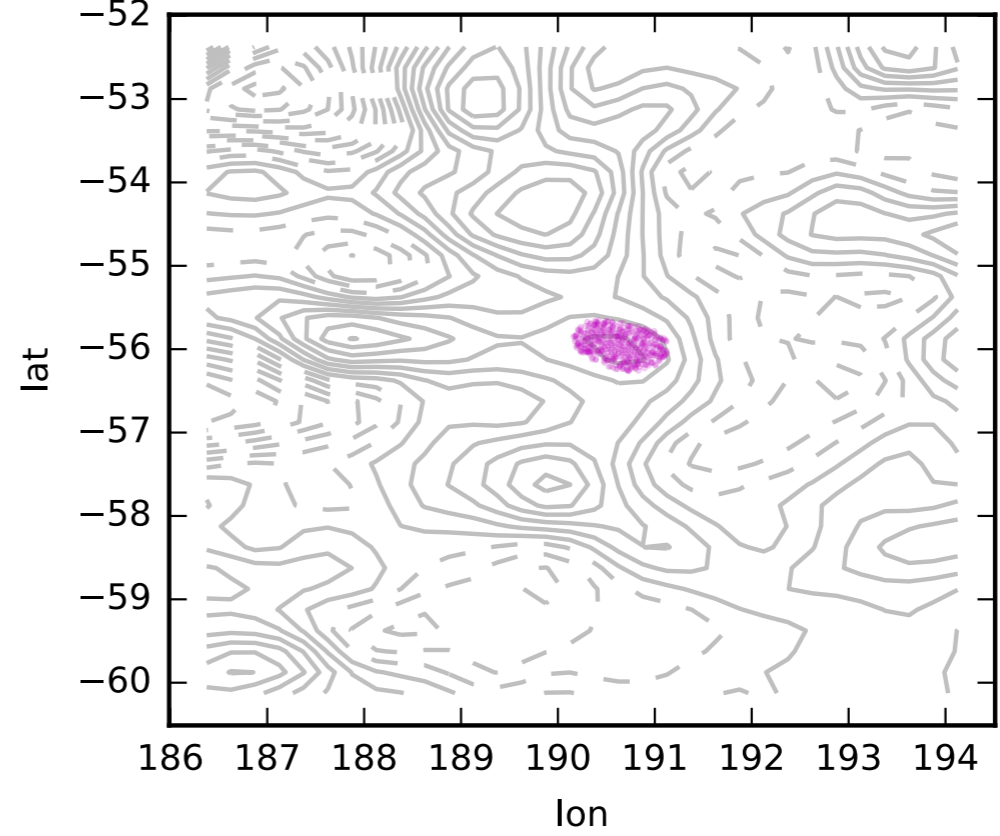
2005-04-28



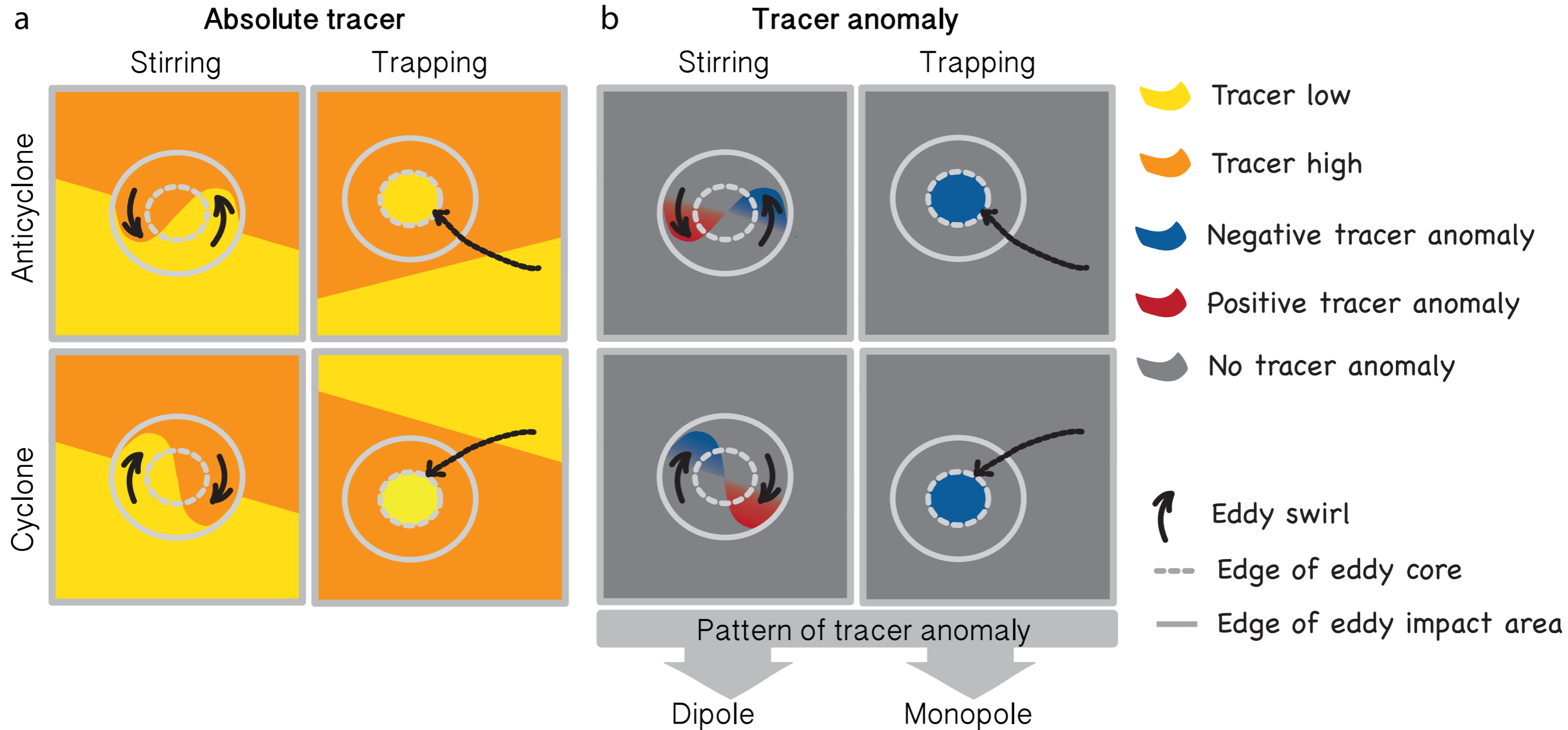
2007-04-18



2007-07-17

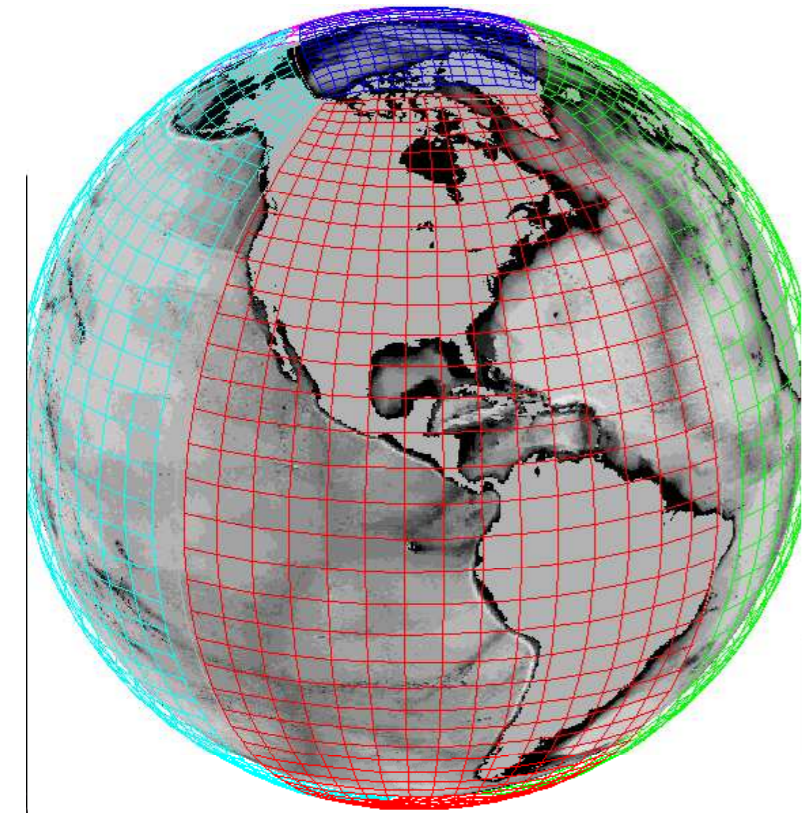


Phenomenology of eddy flux



Frenger et al. (2015)

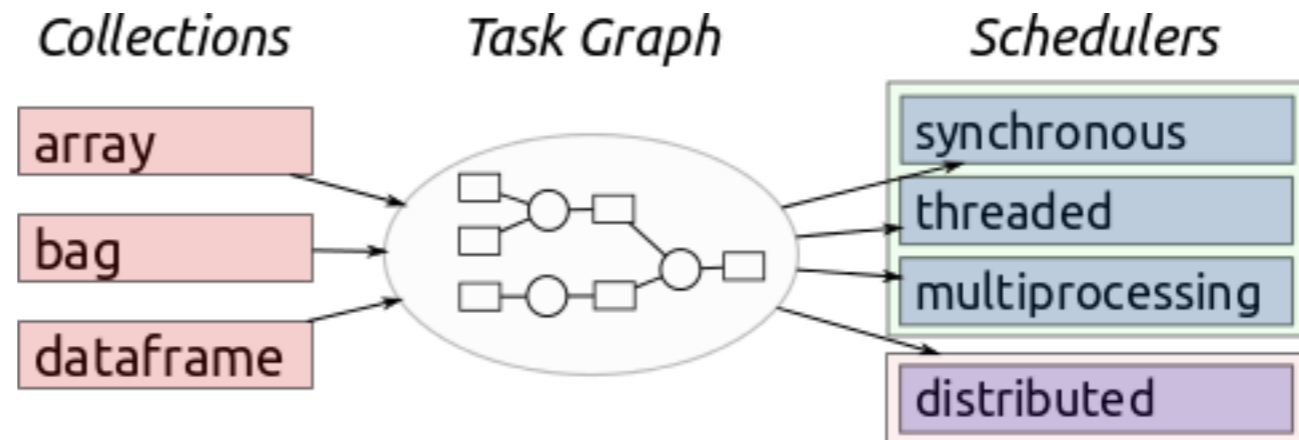
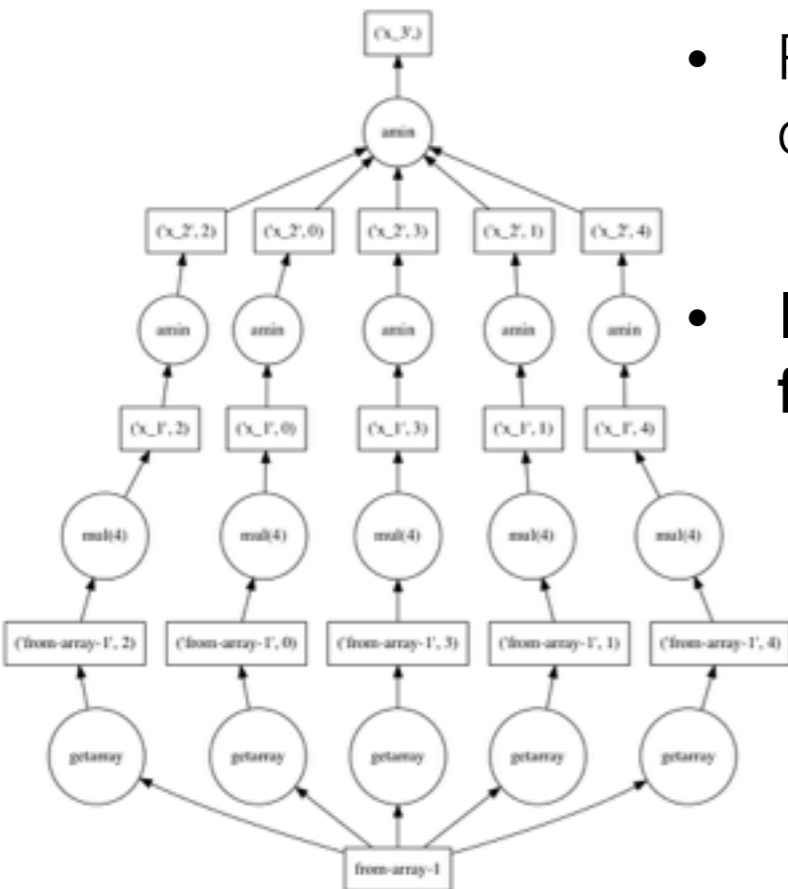
The Tools: MITgcm for Lagrangian Advection



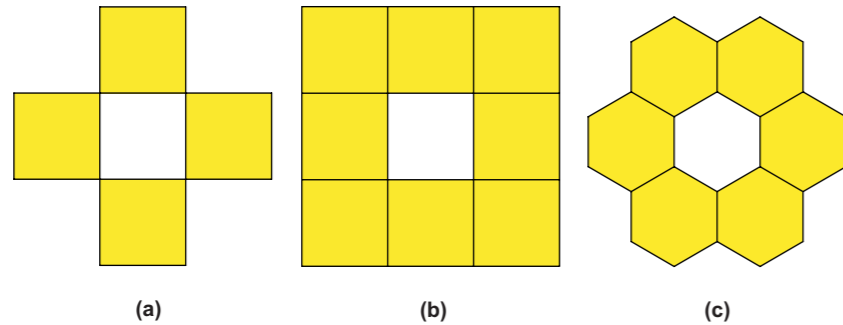
- Has a built in Lagrangian “particle tracking” via FLT package
- Originally designed to simulate ARGO floats
- Full 3D RK4 with trilinear interpolation in space
- “Turbulence” options, but I didn’t use them
- Combined with “offline mode,” which reads velocity fields from files
- Fully MPI parallelized, scales to 100s of processors
- IO bound

The Tools: PyTables and Dask for Post-Processing

- 8 million particles x 5 fields (X,Y,U,V, ω) x 90 days x 32 bits = ~40 GB
- This is “medium data”: fits on a hard drive, but probably not in RAM
- Binary output first converted to **index HDF5** via python PyTables package. Facilitates fast searching and efficient access.
- Processed using **dask**, a parallel task scheduler designed for out-of-core data analysis on multi-core systems
- I put these ingredients together in a lightweight package called **floater**: <https://github.com/rabernat/floater>

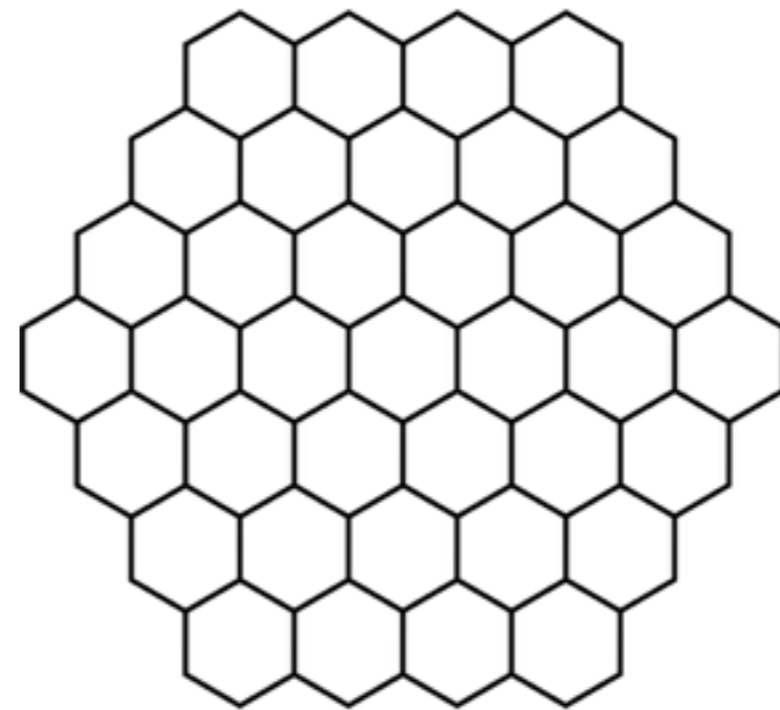


Identifying RLCVs



particles deployed using *hexagonal grid tessellation*
→ unambiguous identification of critical points (extrema and saddles)

- Start from maxima of LAVD
- Add neighboring points until region boundary is no longer convex
- Filter out RLCVs smaller than a certain size threshold (50 km)



Learning from Other Fields

- Global “dense” Lagrangian simulation in high-resolution OGCM would generate ~2 PB of trajectory data
- Compare to “dark sky simulation”
<http://darksky.slac.stanford.edu/>
- “ds14_a is the first trillion particle simulation released to the public, and is being used to study the very largest scales in the Universe. It spans a cubical region 8 Gpc/h on a side, nearly 40 billion light years across. With 102403 (~1.07 trillion) particles, we follow the growth of the largest clusters of galaxies, weighing in at more than a quadrillion times the mass of our own sun.”
- Open data access through <http://darksky.slac.stanford.edu/edr.html>

“Eddy Fluxes” (Eulerian View)

$$\theta = \theta(x, y, t)$$

$$\frac{D\theta}{Dt} = \frac{\partial\theta}{\partial t} + \mathbf{v} \cdot \nabla\theta$$

advection of a scalar
(e.g. temperature)

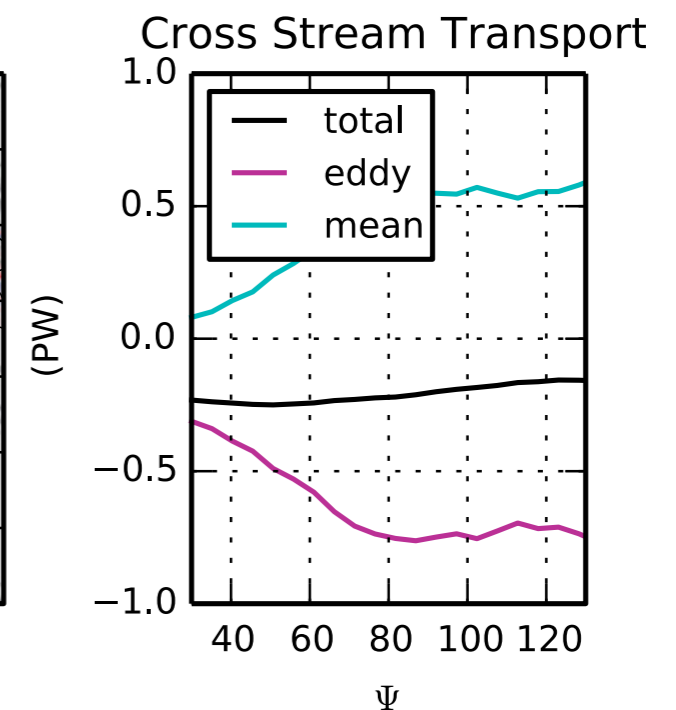
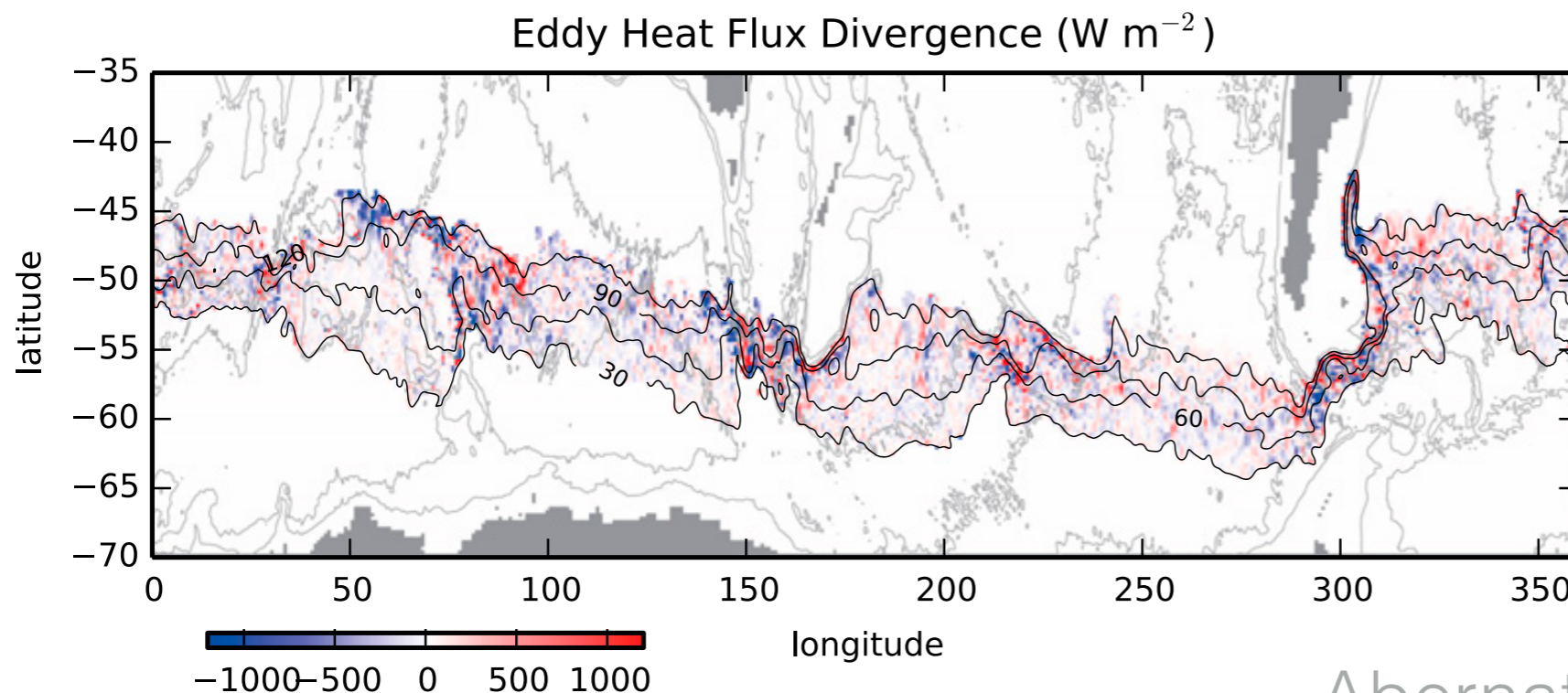
Reynolds average:

$\overline{(\)}$ = time average

$$\frac{\partial\bar{\theta}}{\partial t} + \bar{\mathbf{v}} \cdot \nabla\bar{\theta} = \nabla \cdot (\overline{\mathbf{v}'\theta'})$$

$(\)'$ = fluctuation / “eddy”

↑
“eddy flux”



Abernathy and Cessi (2014)