

Benchmarking Mesoscale variability in global eddy-permitting simulation against satellite data

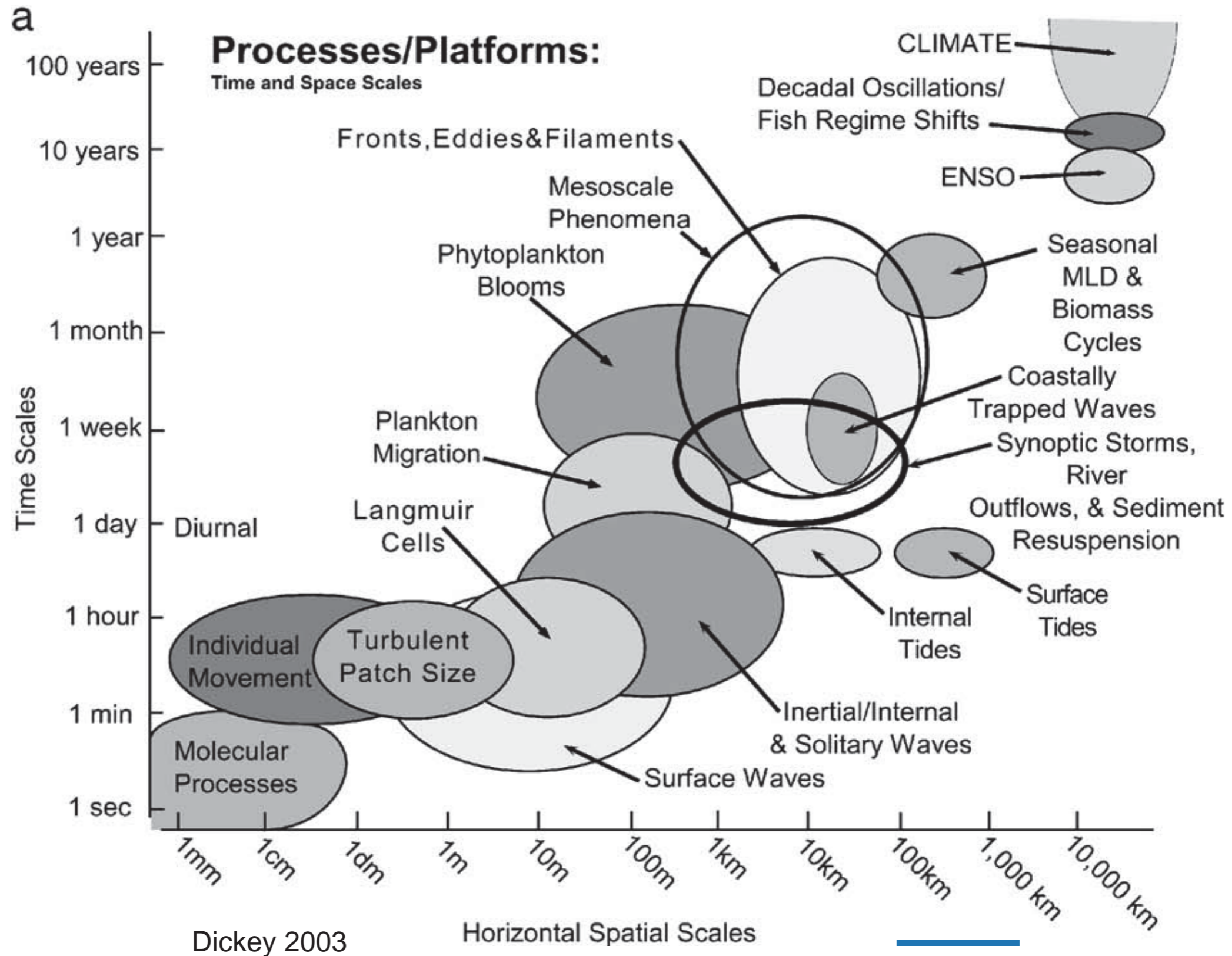
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@NLOA 2016
Madrid

collaborators:
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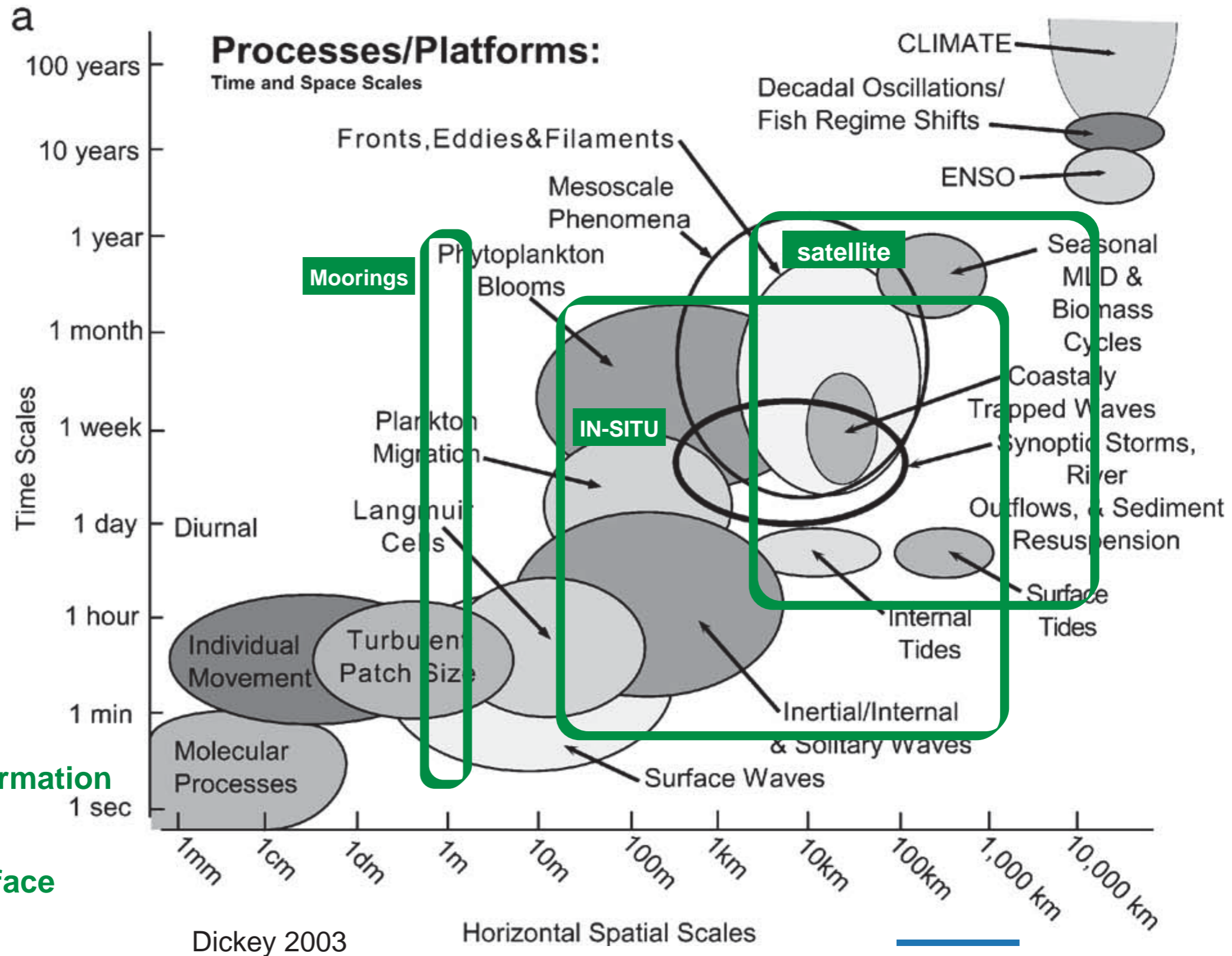


Outline



Eddy permitting models

Outline



Few and sparse information from in-situ data

satellite monitor surface physics only

Eddy permitting models

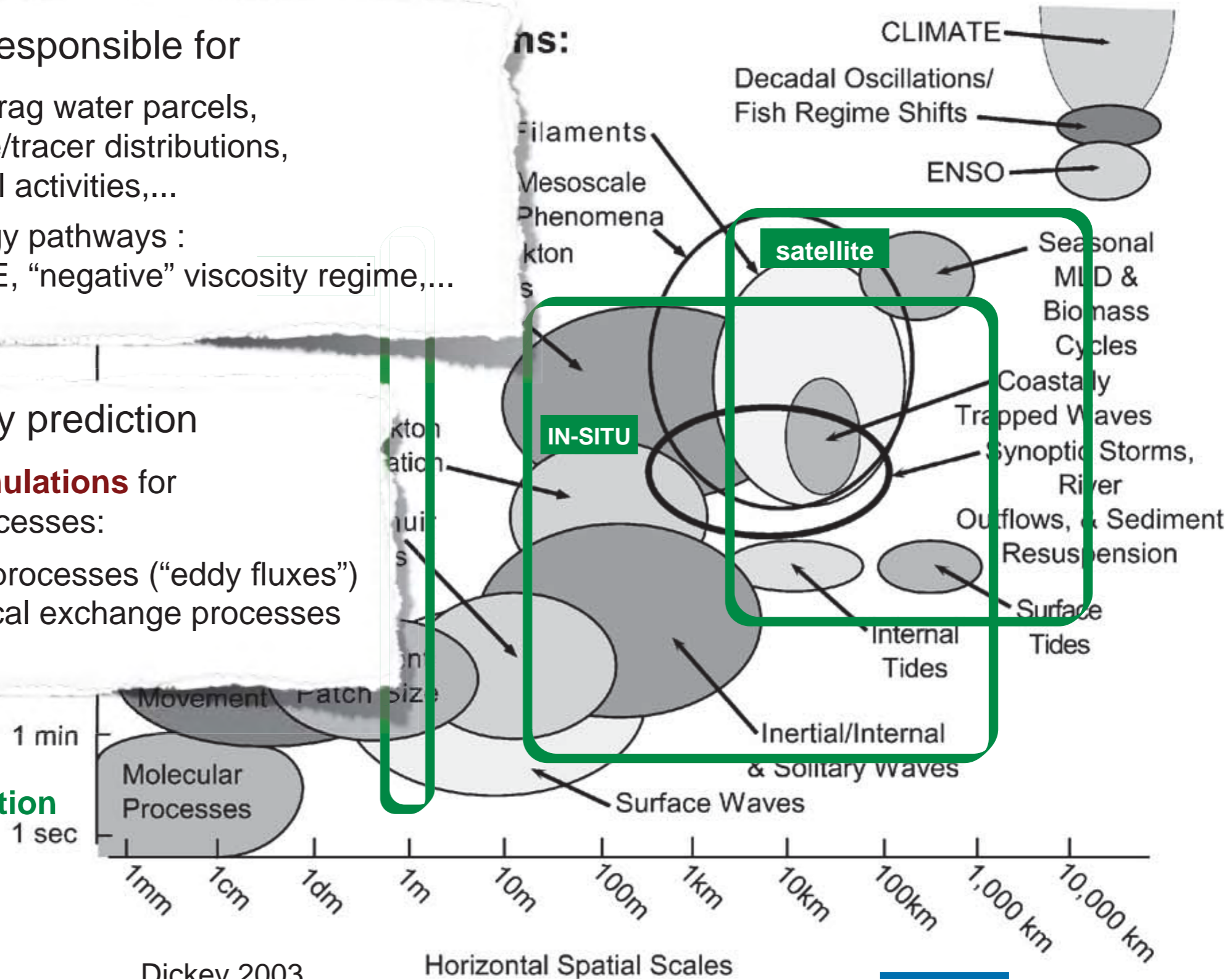
Outline

□ Eddies are responsible for
 as ocean probes : drag water parcels,
 influencing resource/tracer distributions,
 enhancing biological activities,...
 as preferential energy pathways :
 transferring PE to KE, “negative” viscosity regime,...

□ Quantitatively prediction
 need to rely **on simulations** for
 main unknown processes:
 -- lateral exchange processes (“eddy fluxes”)
 -- upper-ocean vertical exchange processes
 (“vertical exchange”)

Few and sparse information
 from in-situ data

satellite monitor surface
 physics only



Dickey 2003

Eddy permitting models

Outline

Role for assimilation :

- correct initial ocean state towards a more realistic description
- include possible missing features caused by poor resolution



at any length-scale? What's its performance at mesoscale level?

Quantitative assessment of variability over three datasets

1) NEMO free simulation
at 1/4
10y run 2003-2012

2) C-GLORS reanalysis of NEMO
simulation at 1/4
10y 2003-2012
(Storto et al. 2014)

3) Satellite/scatterometer
dataset at 1/4
10y run 2003-2012
(AVISO/OSCAR datasets)

- Global ocean basin

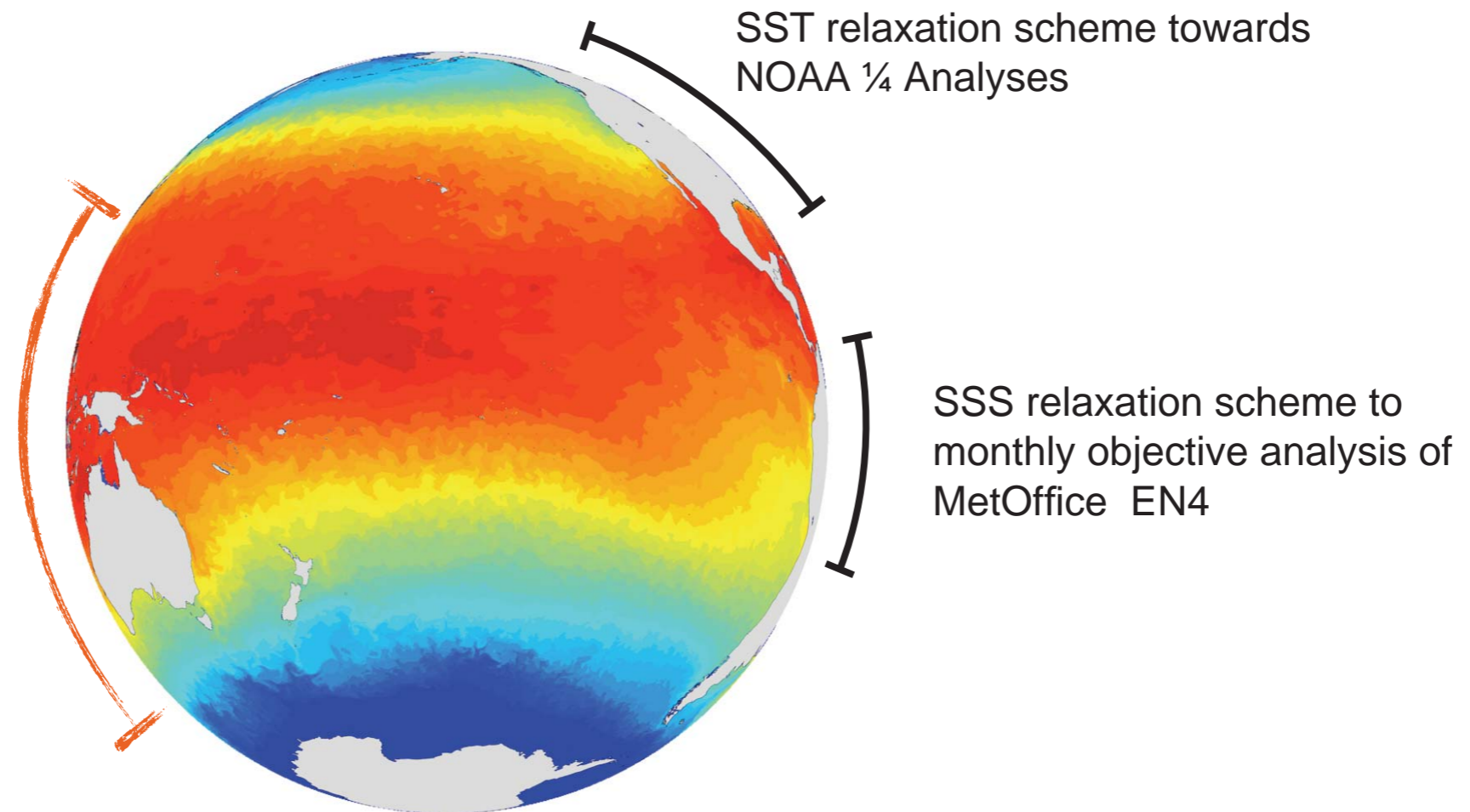
OceanVar assimilation scheme

- C-Glors exploits a 3Dvar assimilation scheme (*Storto et al. 2015*) with updates from multiple datasources.

3Dvar assimilation data :

- in-situ temperature and salinity (XBT, CTD, Argo, Moorings, etc. assembled by Ifremer)

- along-track satellite altimetry observations (Jason-2, Altika and CryoSat2 by CLS/AVISO)



Corrections are added to the surface/sub-surface salinity and temperature (Cooper et al. 1996)

A plethora of schemes in literature

Regional eddy tracking

- Okubo-Weiss Method *Weiss (1991), Isern-Fontanet et al. (2004)...*
- 2D-Wavelet Methods *Doglioli et al. (2007), ...*
- Winding angle Methods *Sandarjoen et al. (2000), ...*
- Geodesic transport theory *Beron-Vera et al. (2013), ...*
- Geometric analysis UV fields *Nencioli et al. (2010), ...*
-

Some comparison with in-situ or satellite data

Trani et al. (2011), Chaigneau et al. (2011), Griffa et al. (2008), Shoosmith et al. (2005)...



Methods for Global Ocean

□ This overall census shrinks to a bunch of methods considering the global ocean....

☑ Okubo-Weiss method

local balance between deformation and vortical flow

$$W = \underbrace{(v_x + u_y)^2 + (u_x - v_y)^2}_{\text{strains...}} - \underbrace{(v_x - u_y)^2}_{\text{relative vorticity}} < 0$$

Very noisy, tendency of generating false-positive eddies

require a regional cutoff value
 $< \sigma(x, y)$

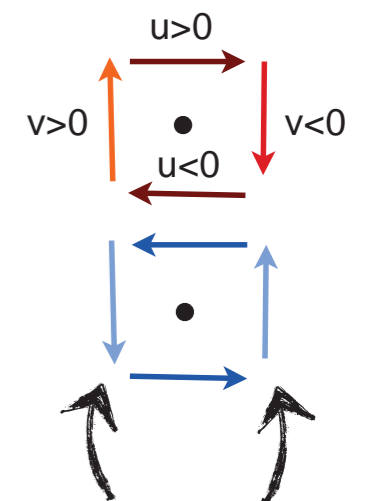
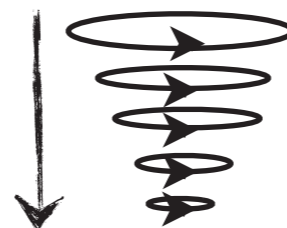
☑ Our method is a modified geometry based algorithm

➔ detection of anomaly SSH pattern (horizontal extension)

➔ detection of rotation inside the eddy at least 5m deep

filter-out spurious eddy-like patterns that do not show a corresponding vortex (close to coastline, merging AVISO SSH and OSCAR UV, ...)

Access to vertical extension of eddies!!!



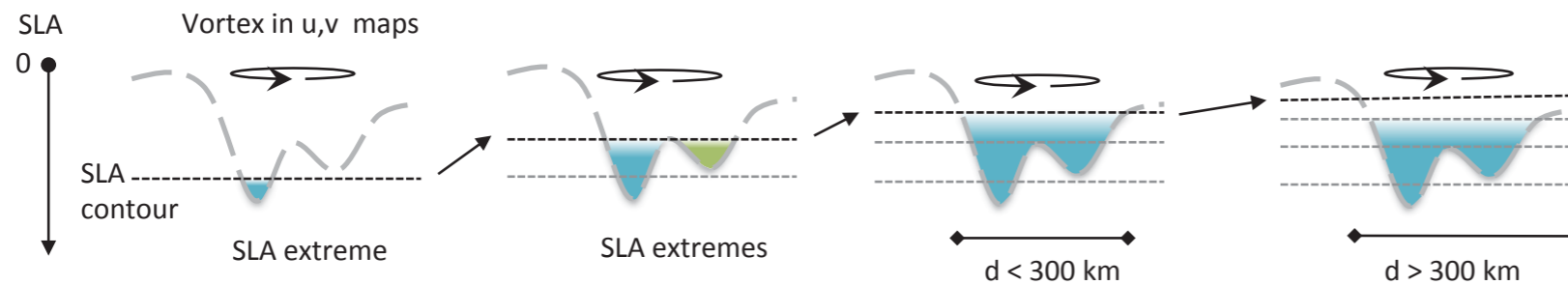
nearest points on a square

Geometry-based method on SLA

□ Horizontal extension emerging from consecutive cuts of SLA contours

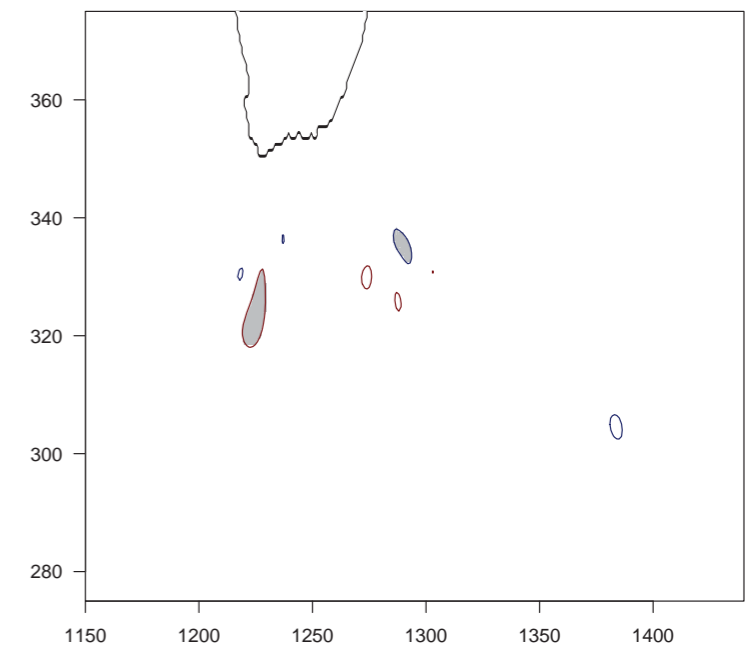
☑ Eddies are limited to a maximum area of d radius

1° method (size)



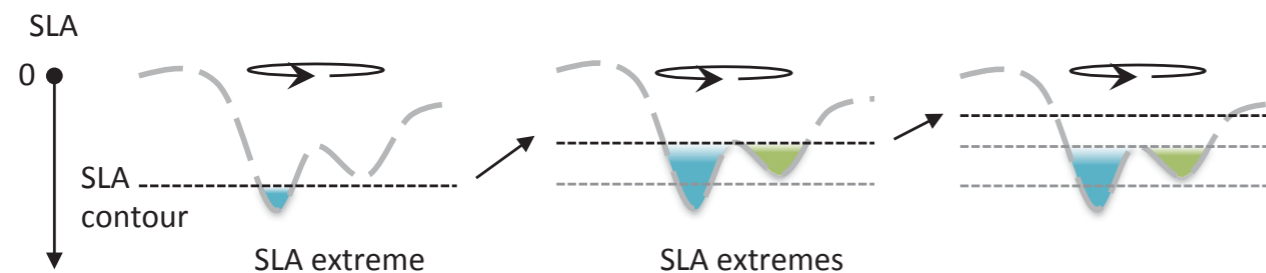
Cons: depending on a parameter d , weak dependency

SLA > 50 cm

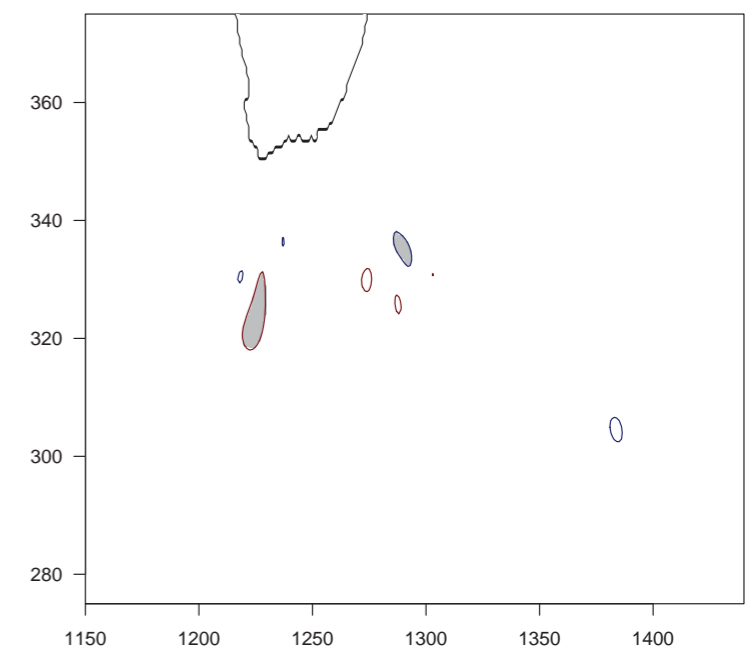


☑ Any local extreme of SLA corresponds to a different eddy

2° method (single extreme)



Cons: double counting “moving” eddies in maps that are averaged over several days.
underestimate of eddy size

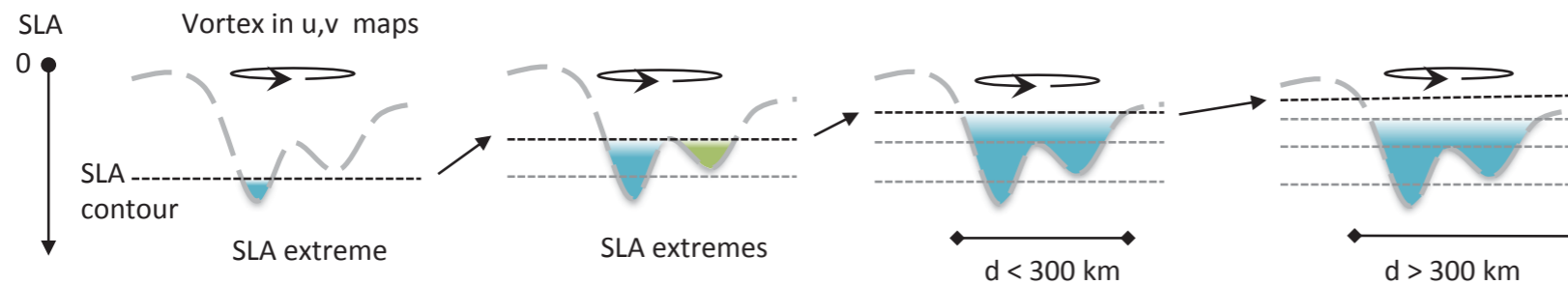


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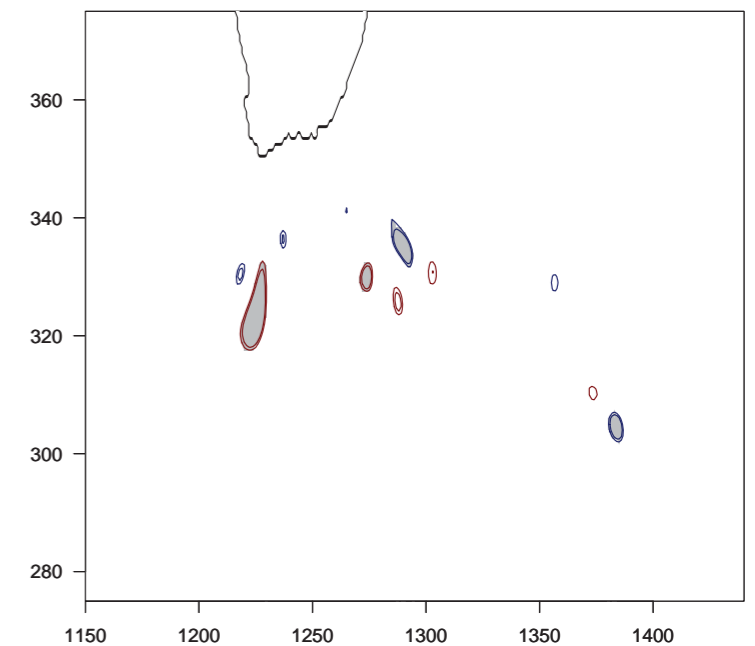
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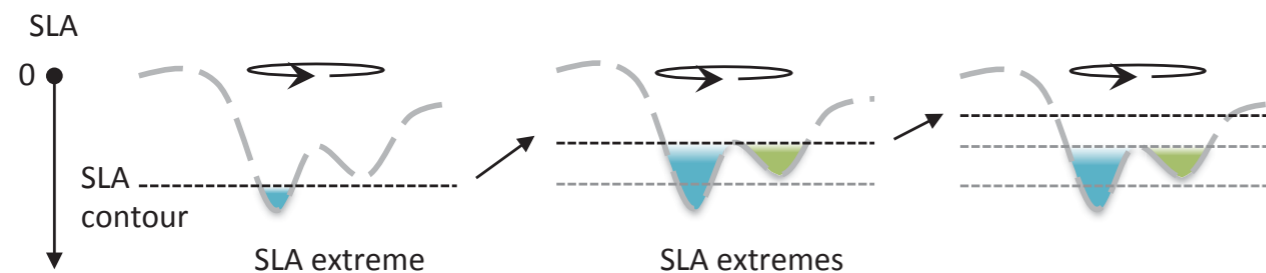
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SLA > 45 cm

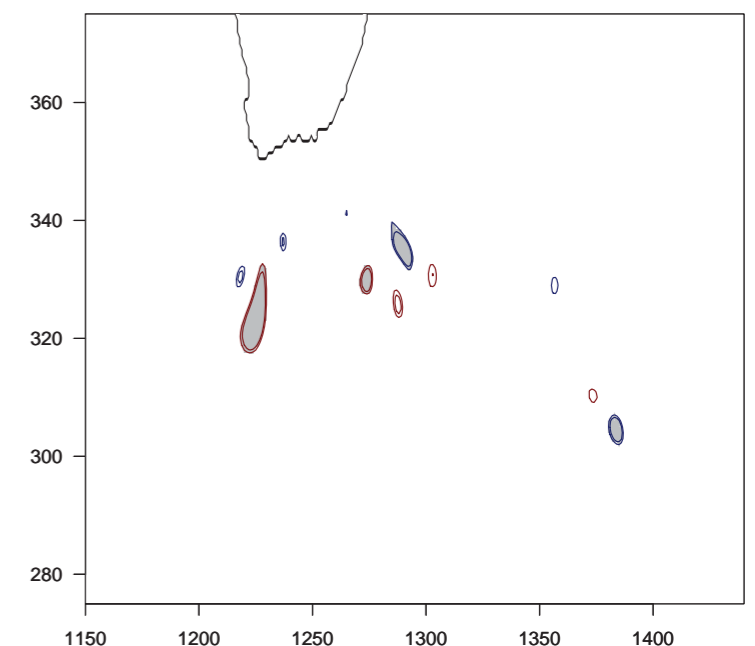


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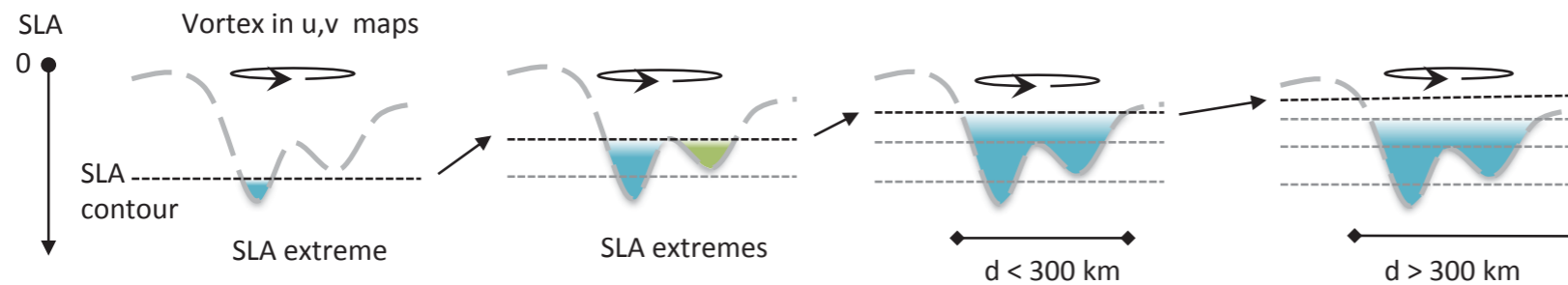


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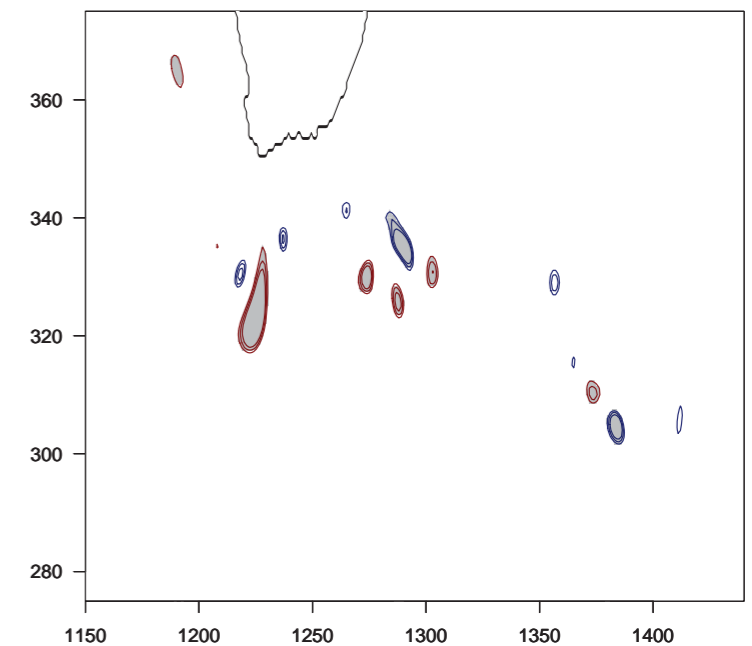
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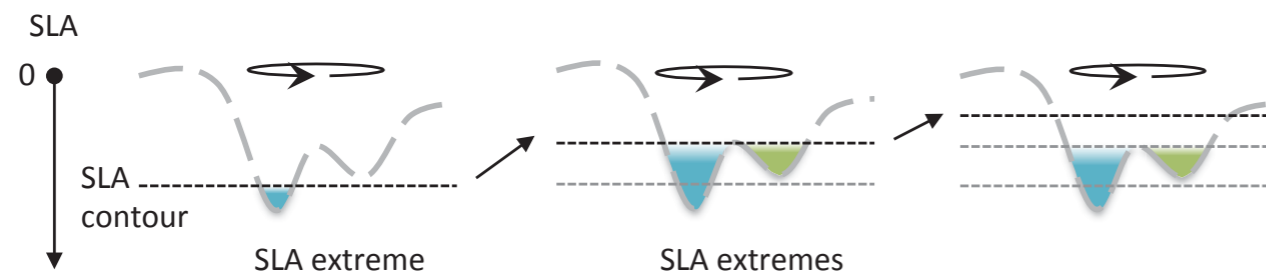
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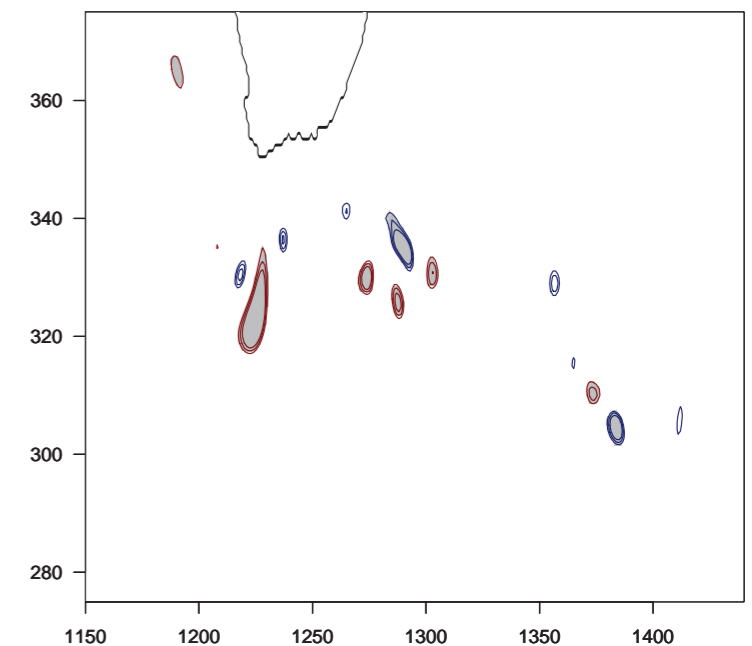


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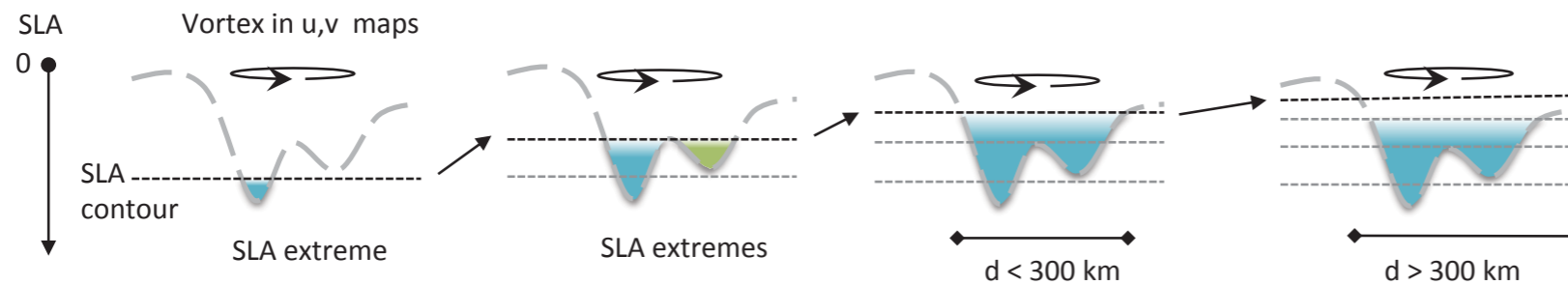


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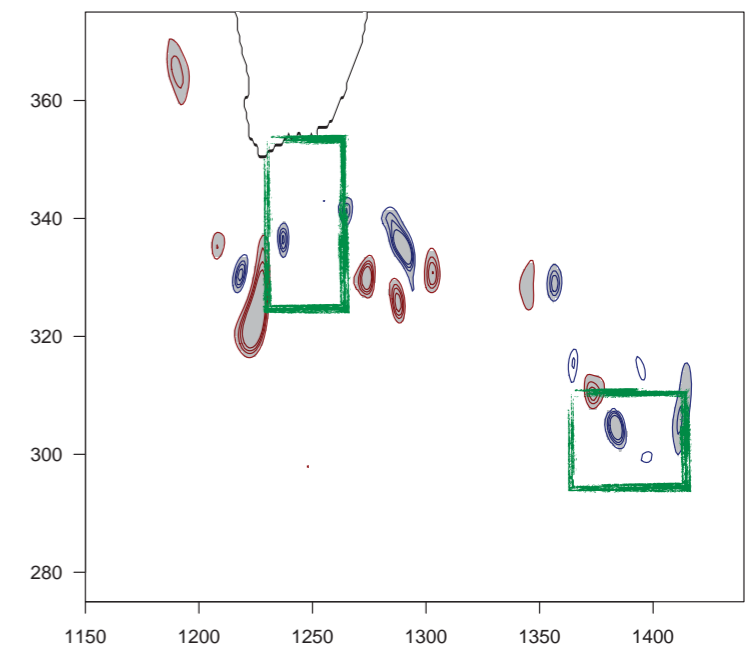
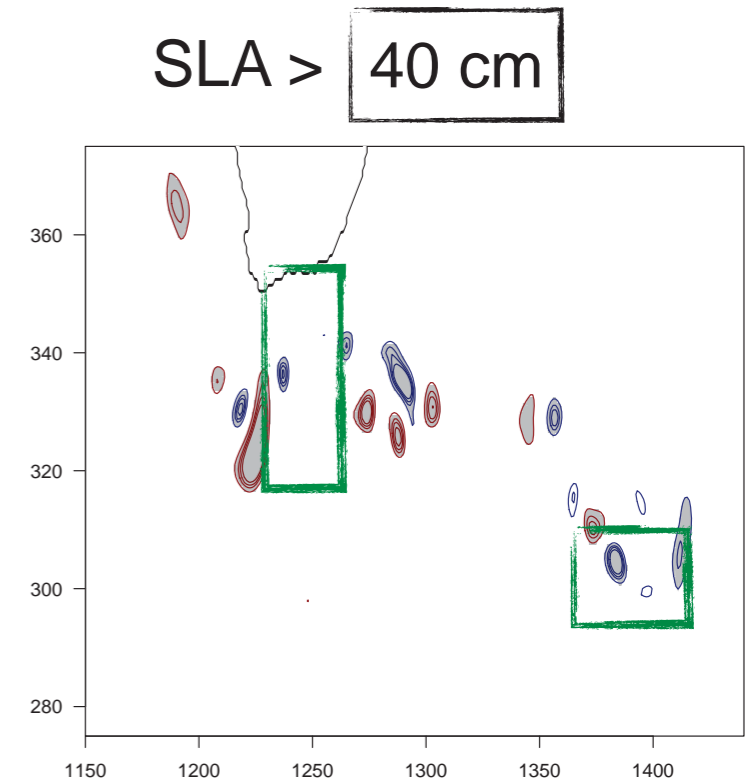
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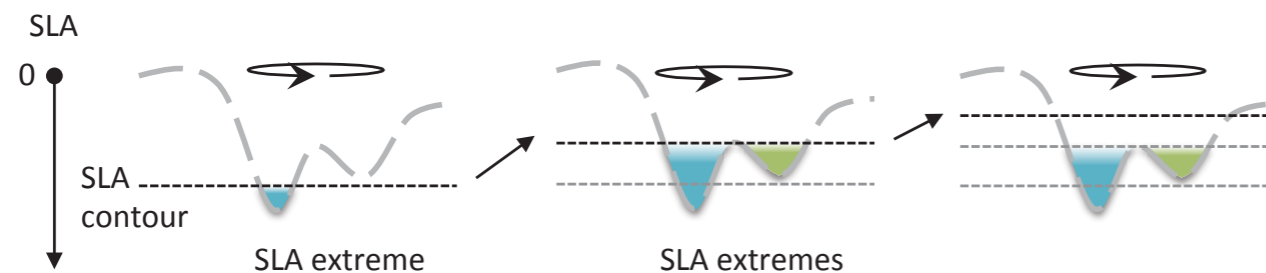


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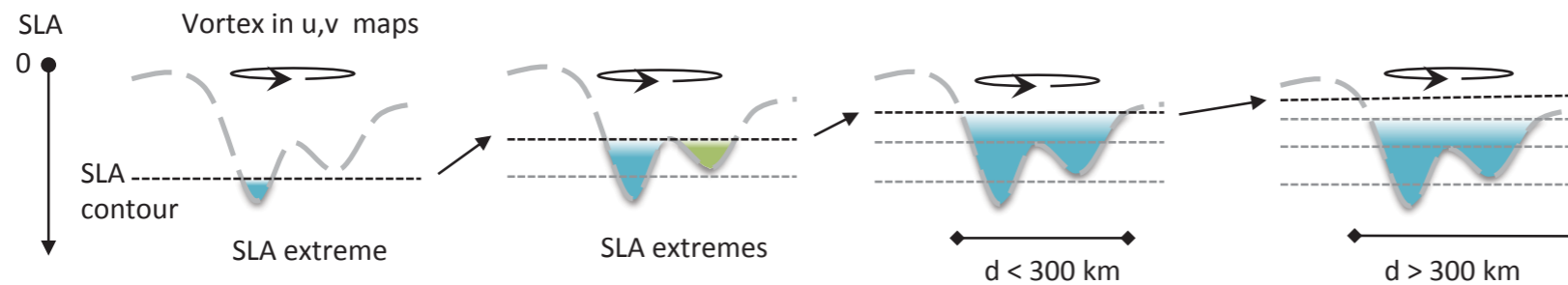
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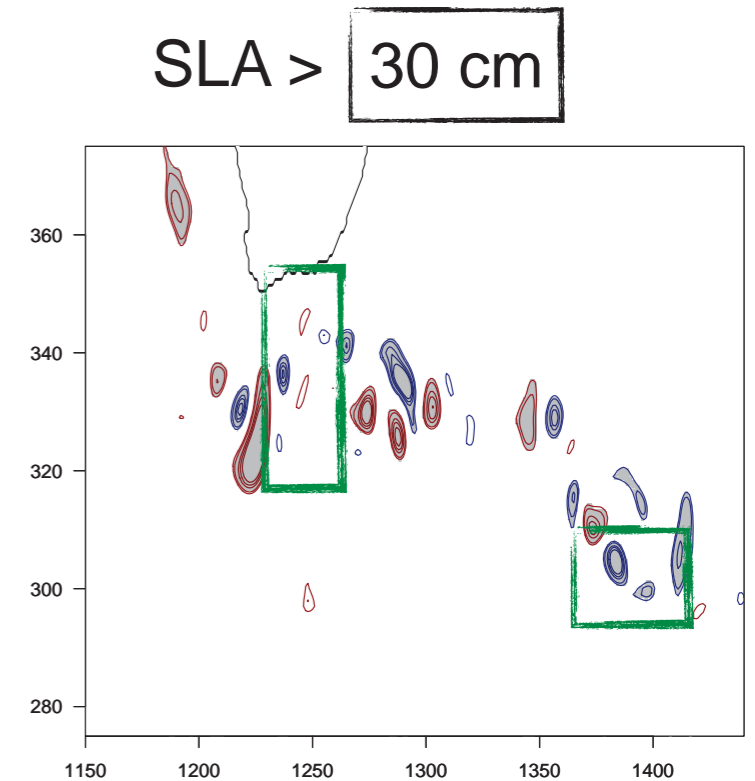
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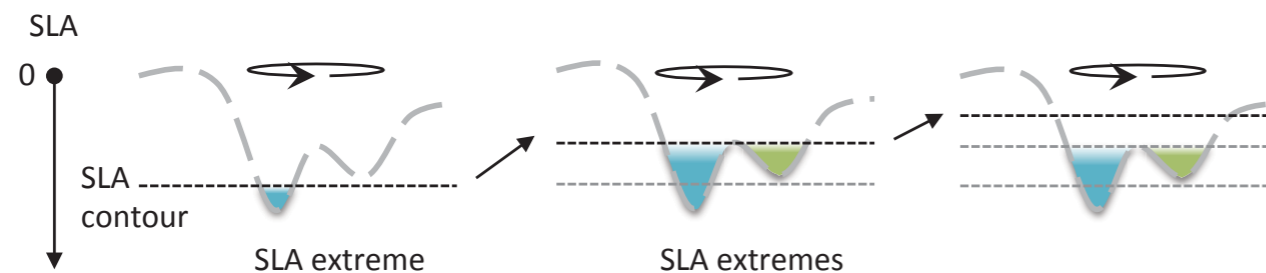


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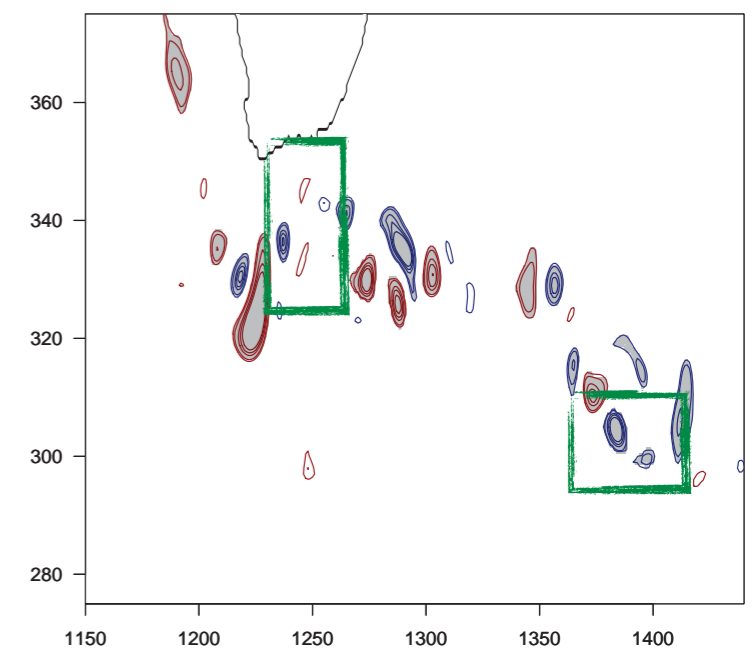


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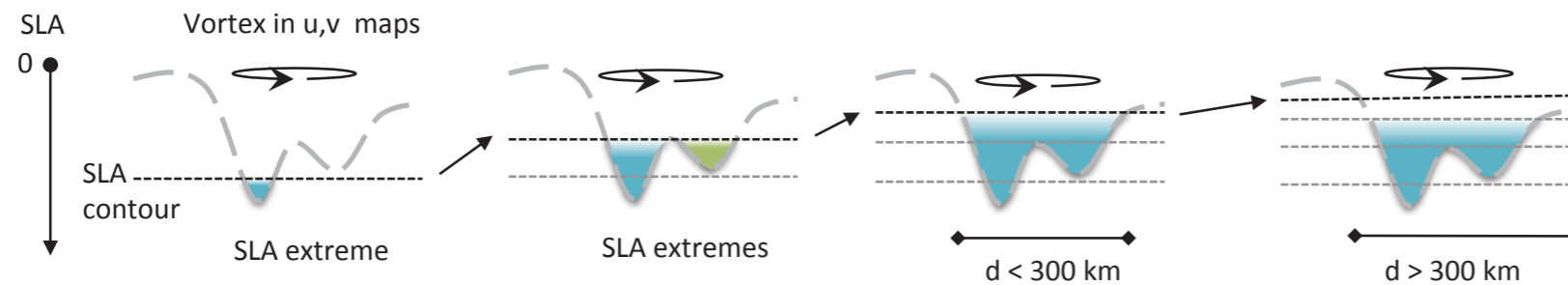


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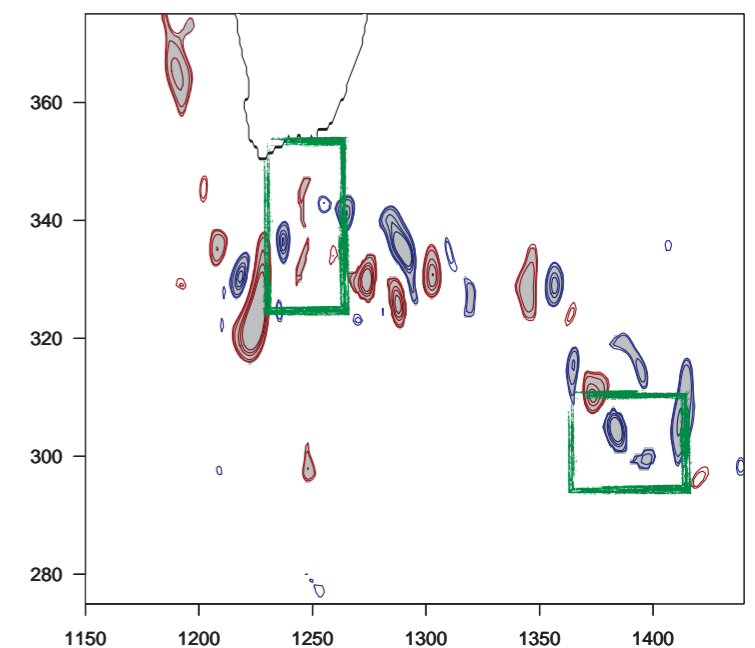
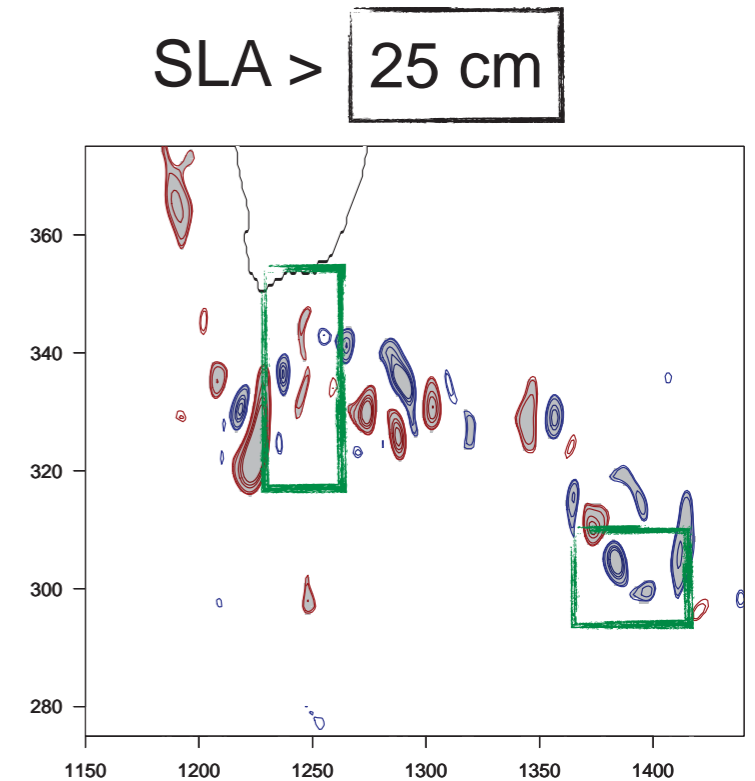
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1° method (size)

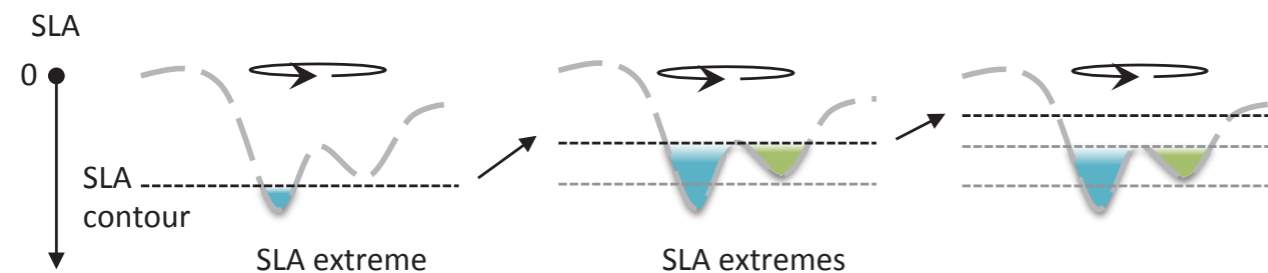


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2° method (single extreme)



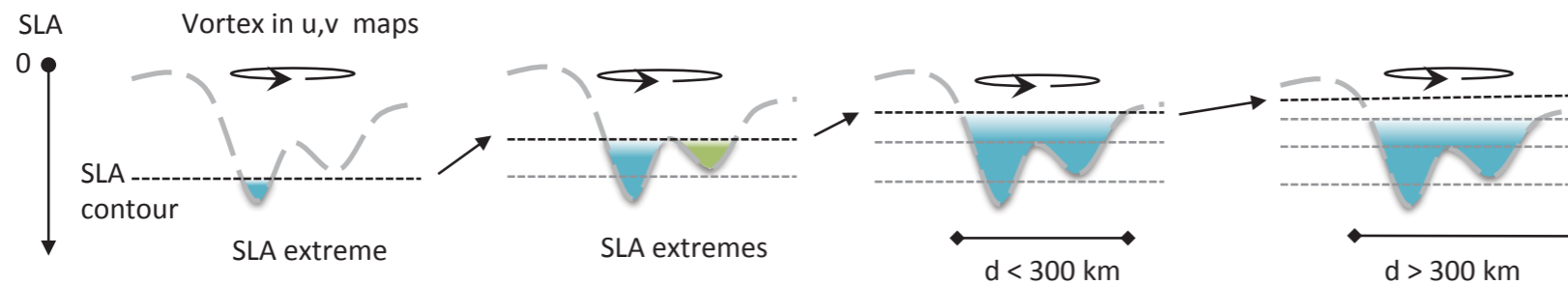
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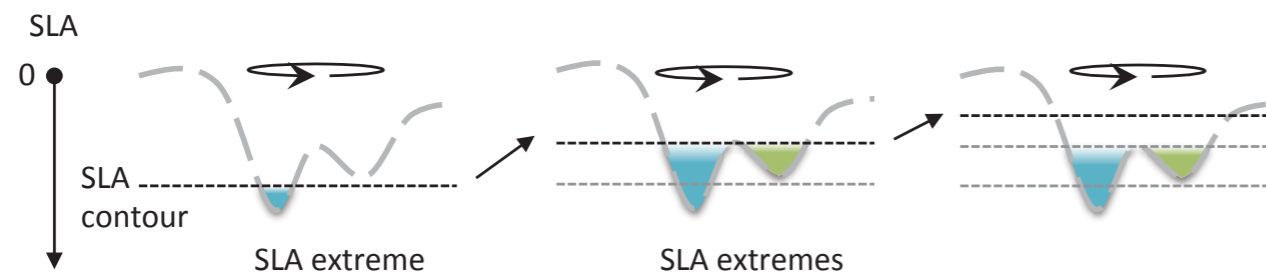
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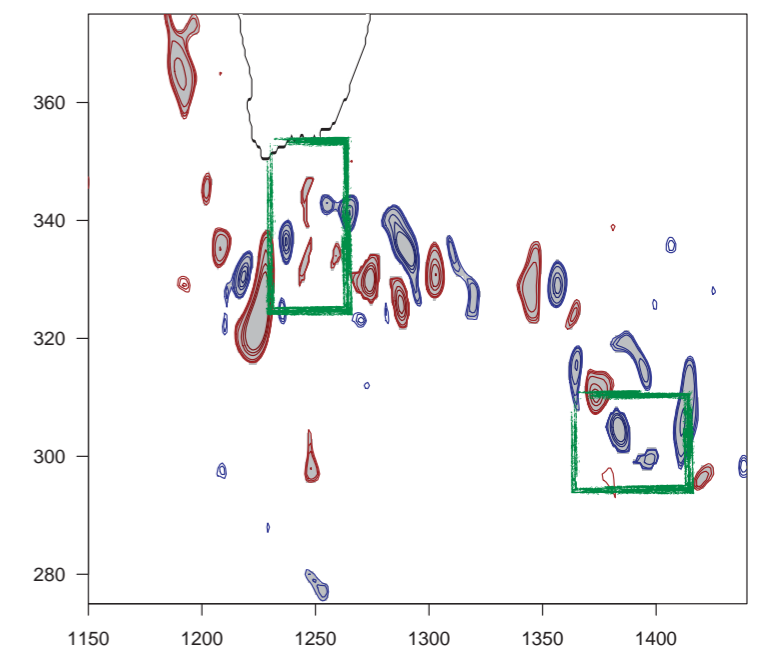
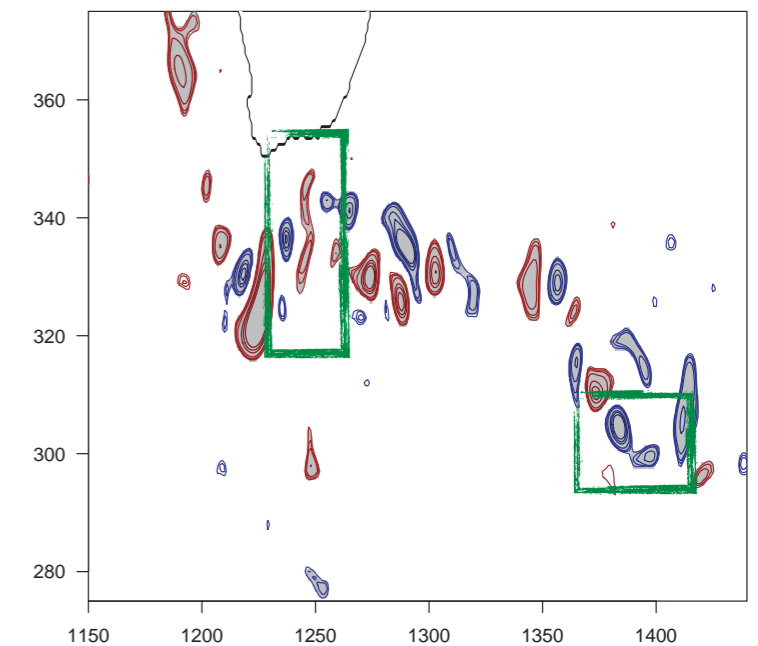
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SLA > 20 cm

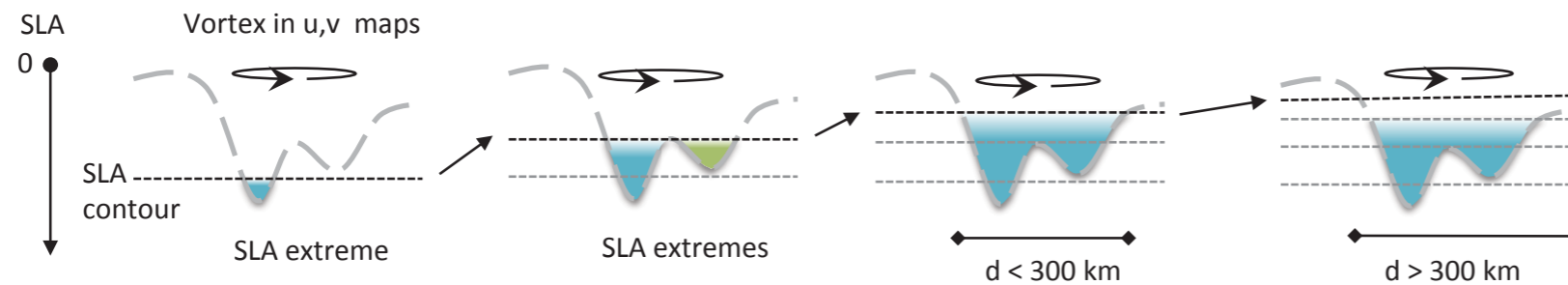


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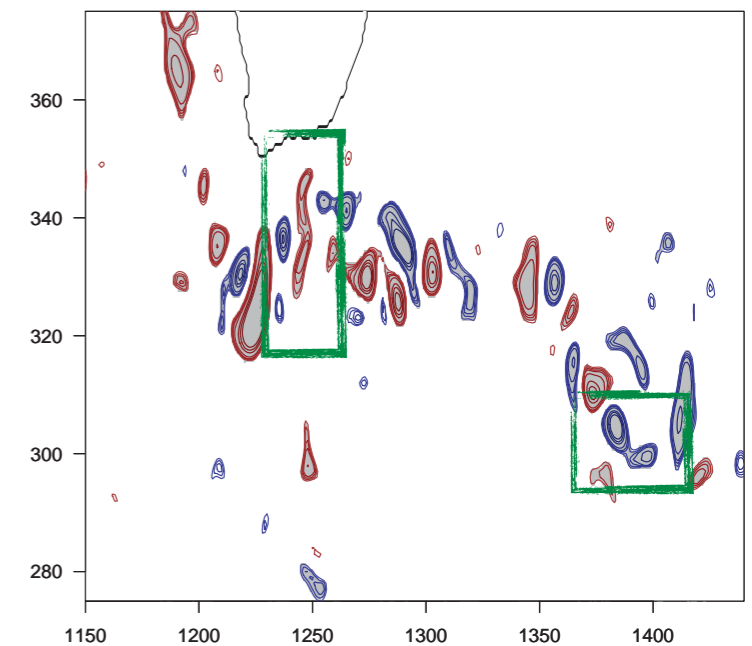
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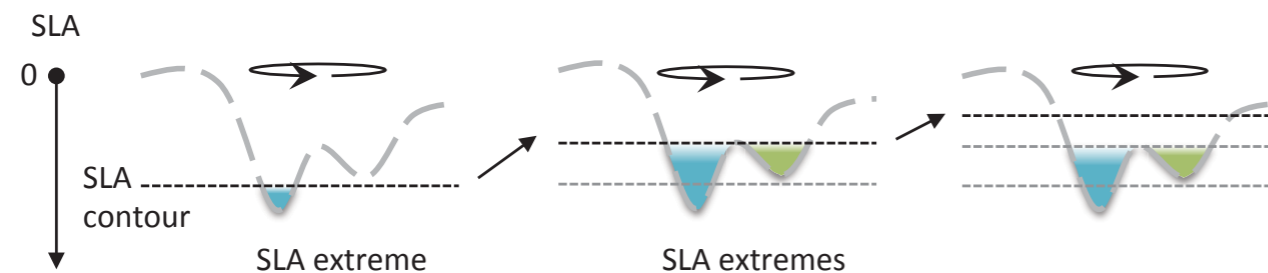
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SLA > 15 cm

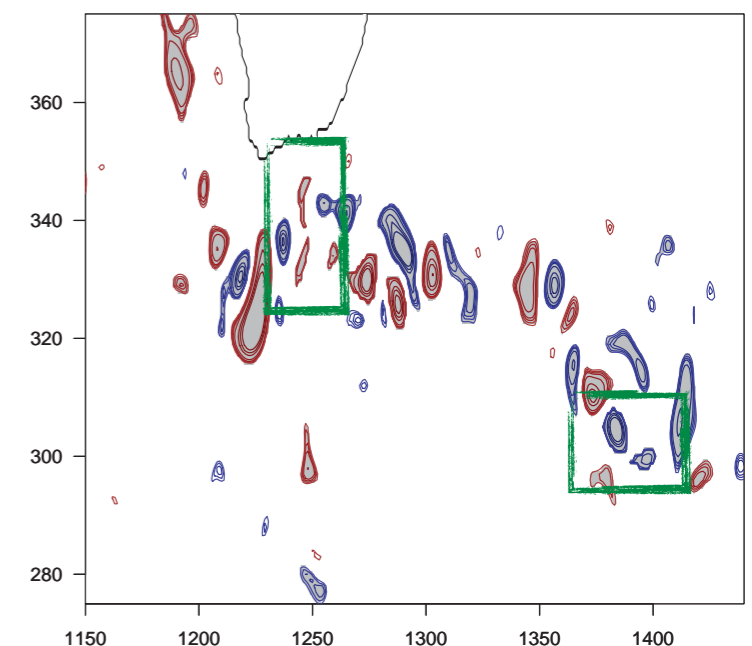


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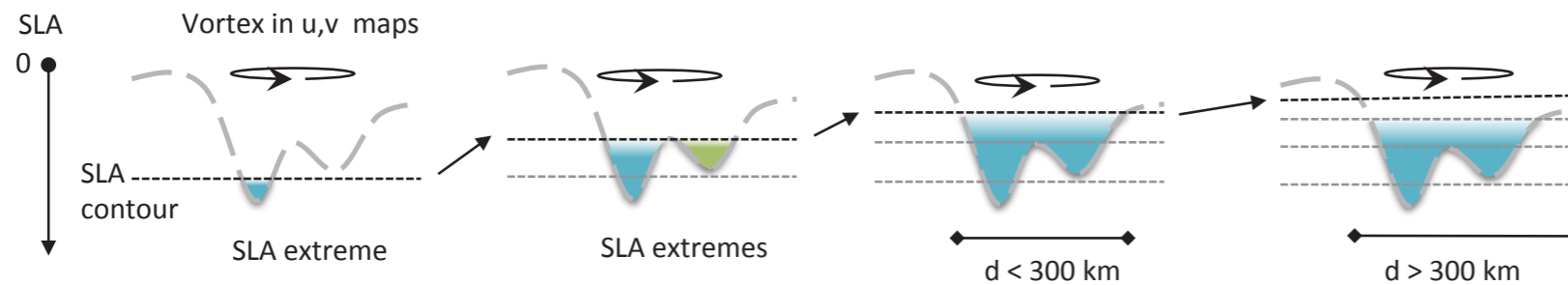


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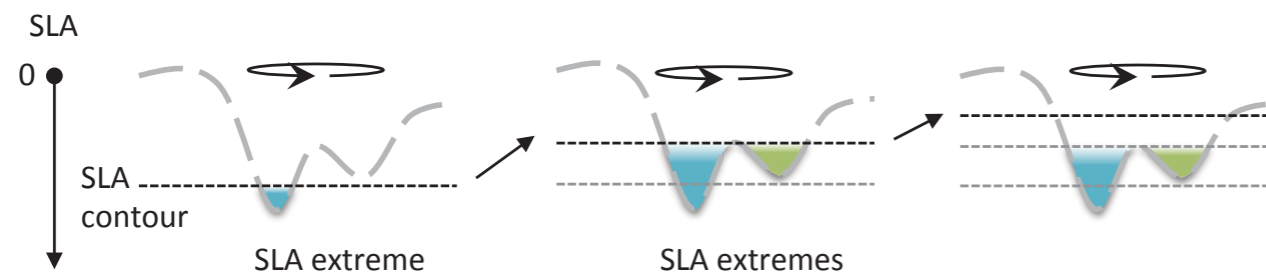
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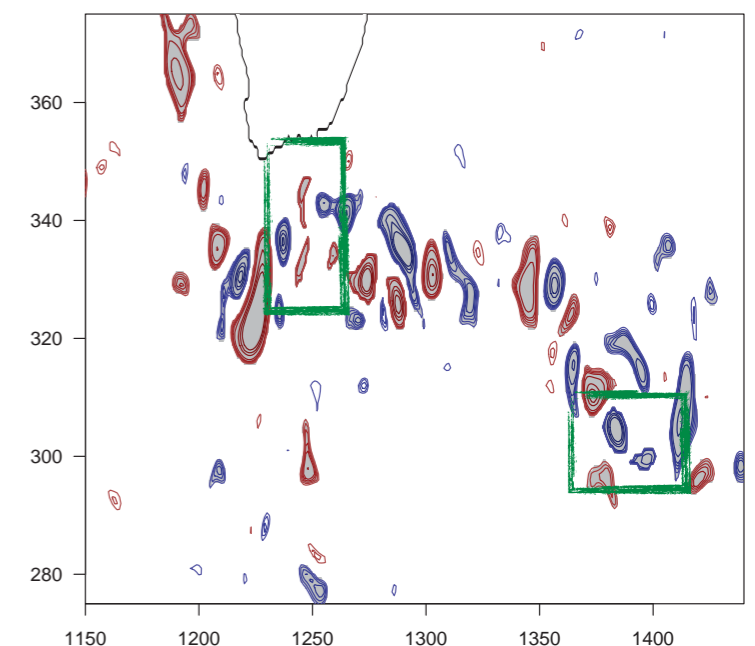
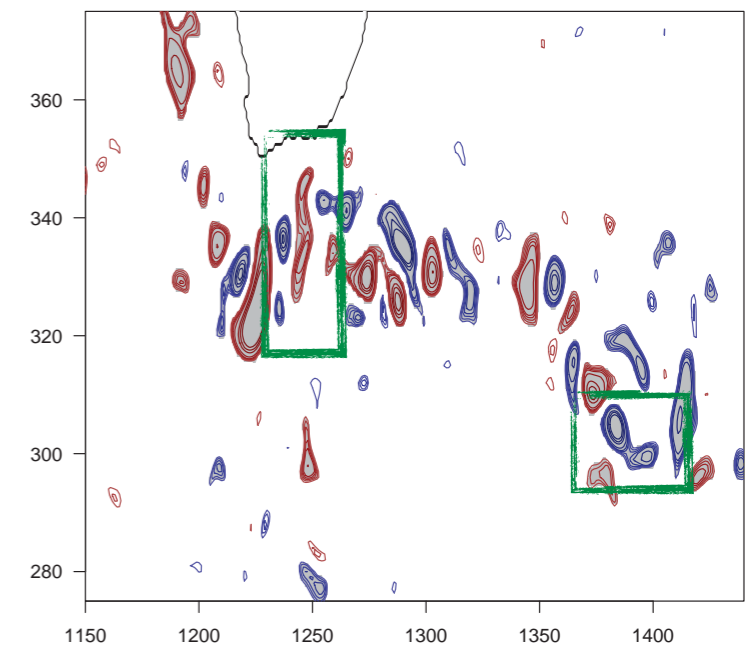
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SLA > 14 cm

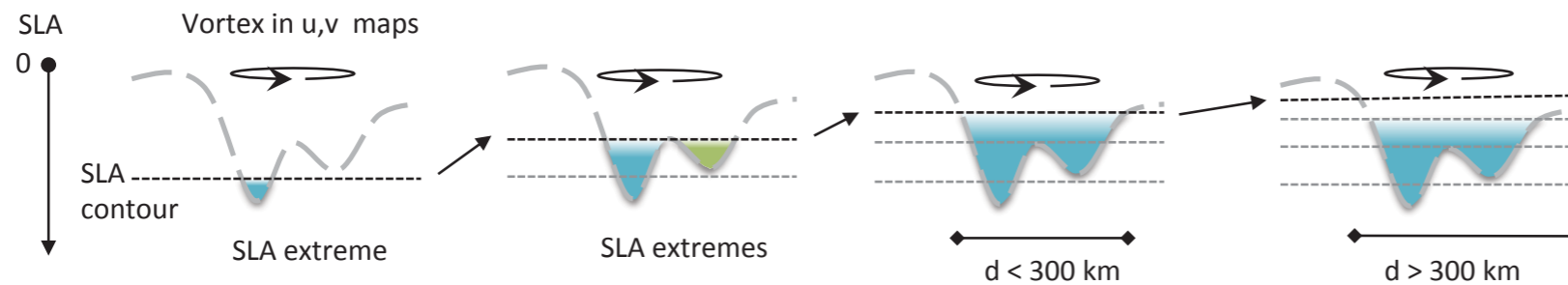


Geometry-based method on SLA

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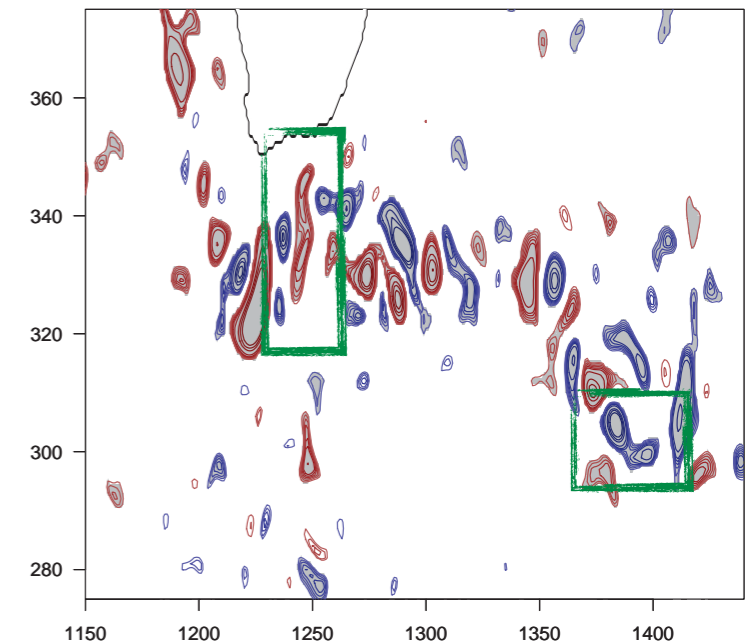
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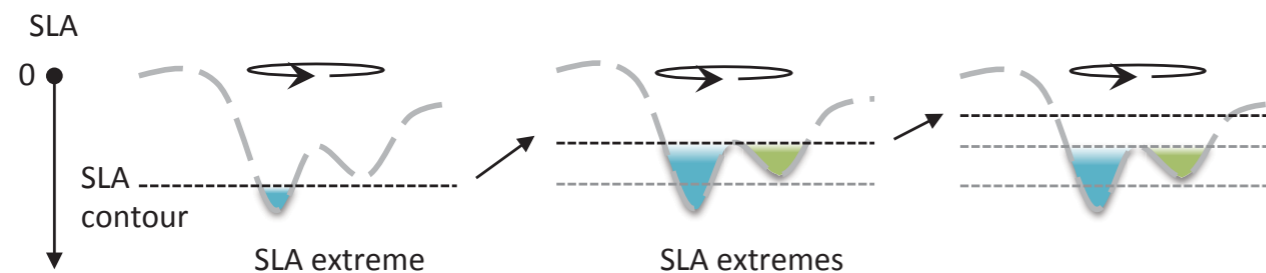
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SLA > 13 cm

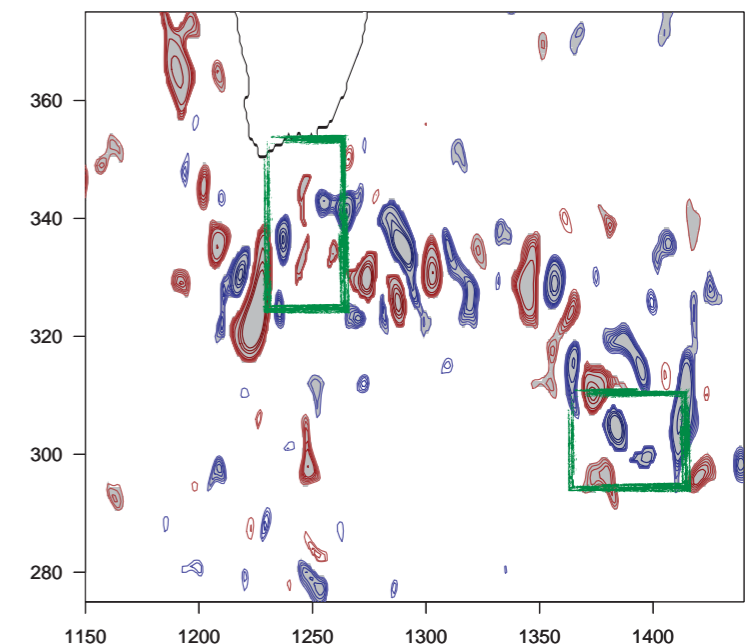


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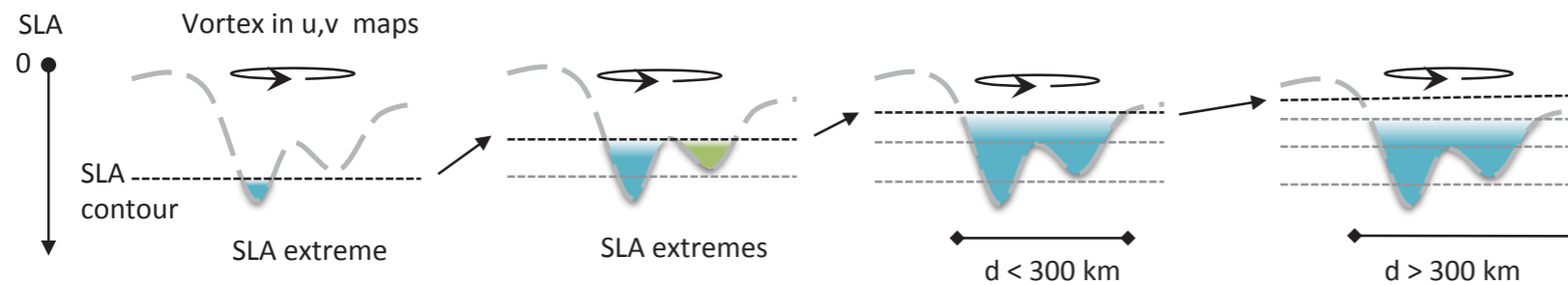


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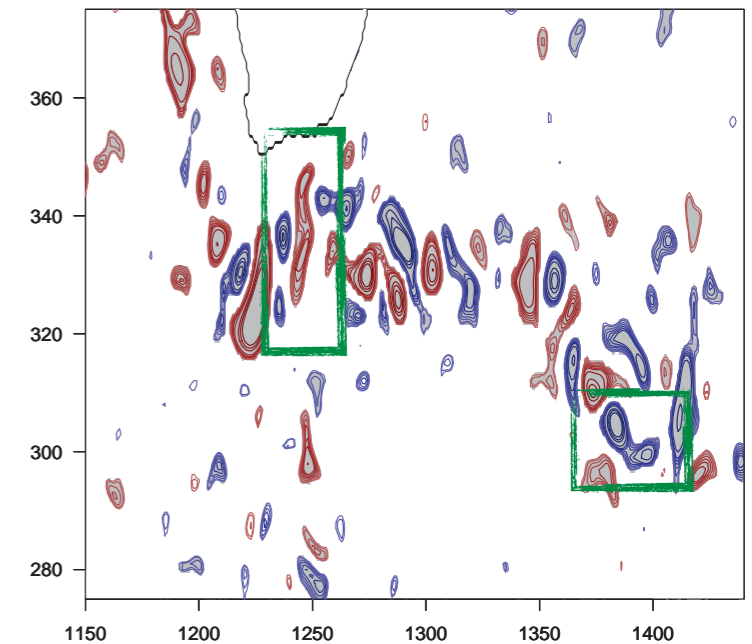
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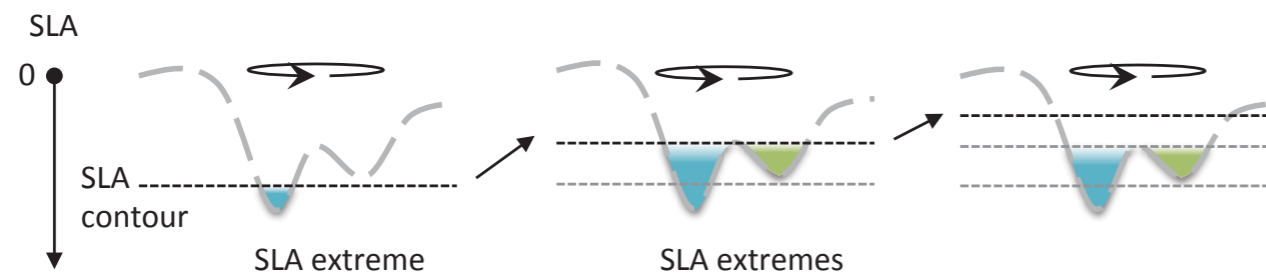
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SLA > 11 cm

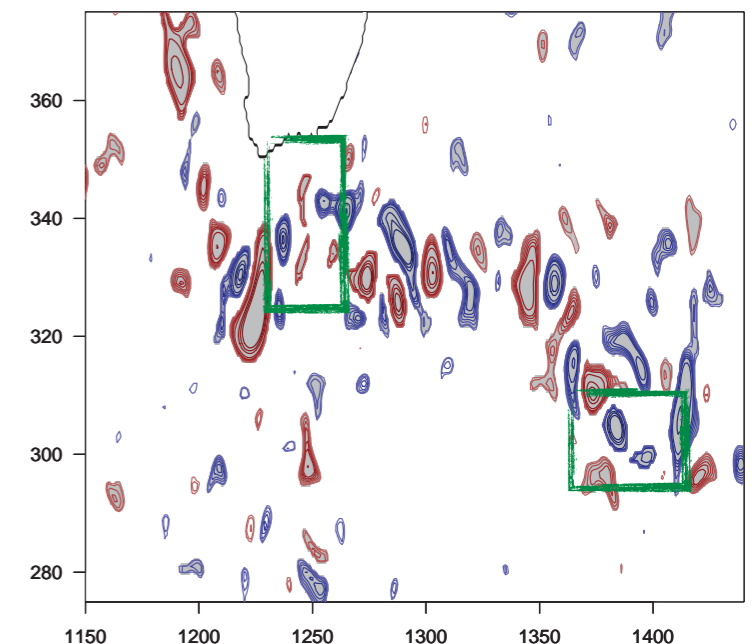


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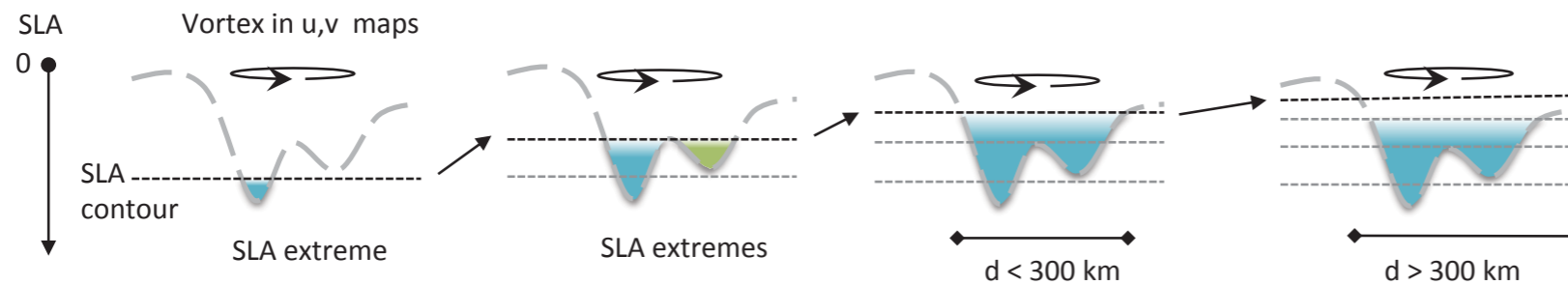


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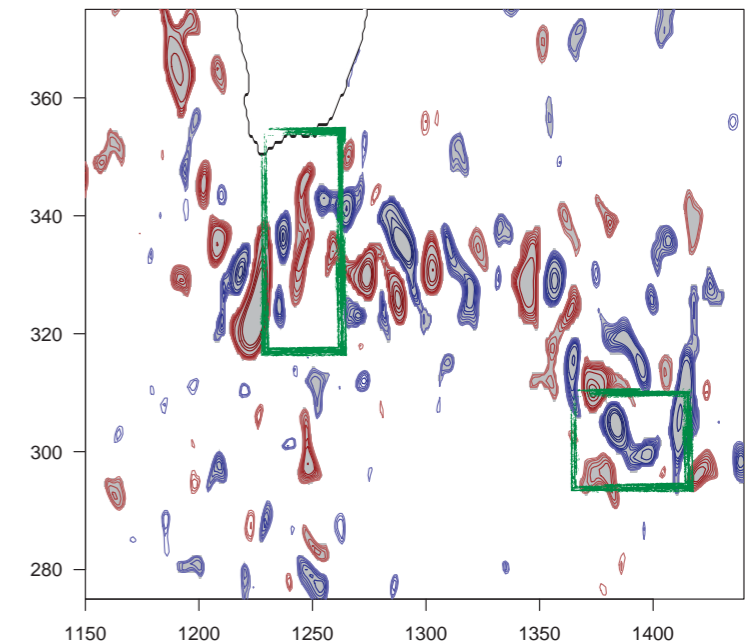
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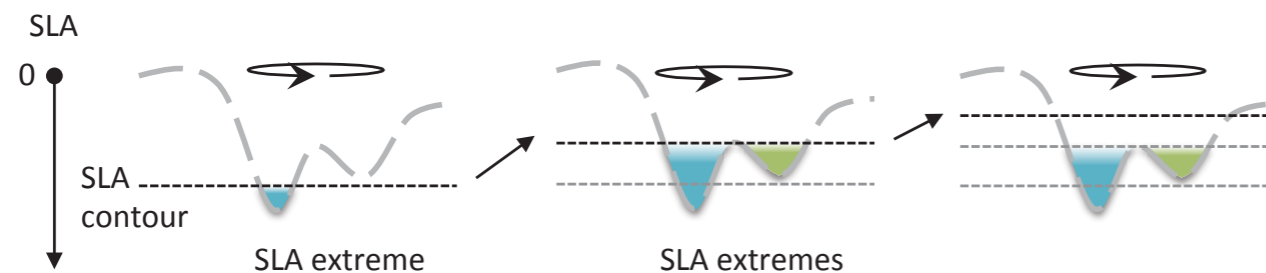
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SLA > 9 cm

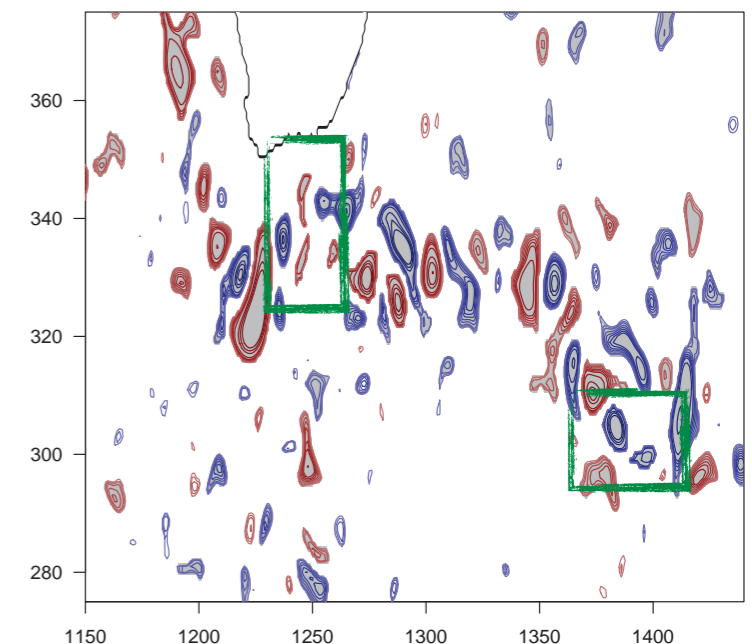


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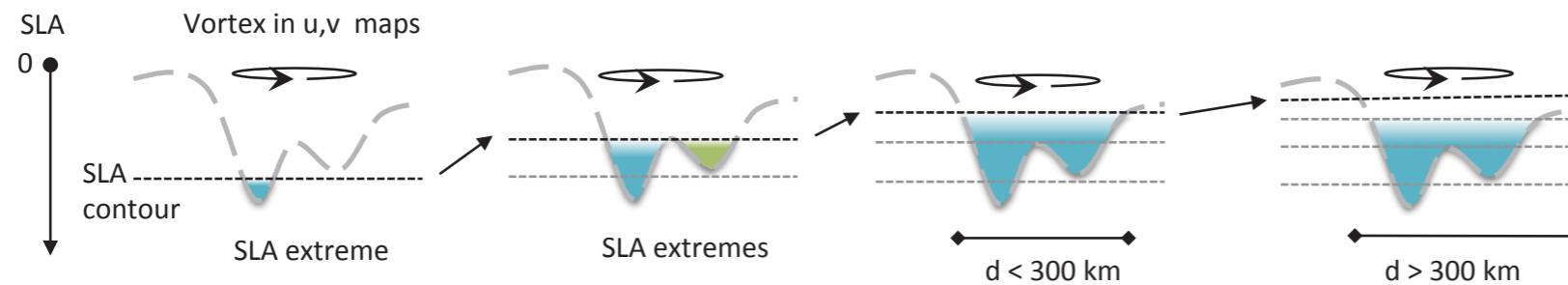


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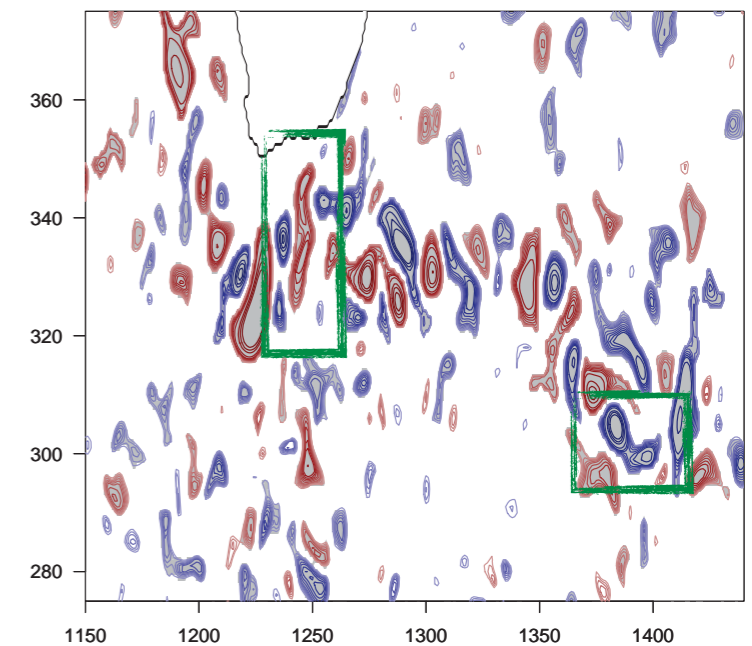
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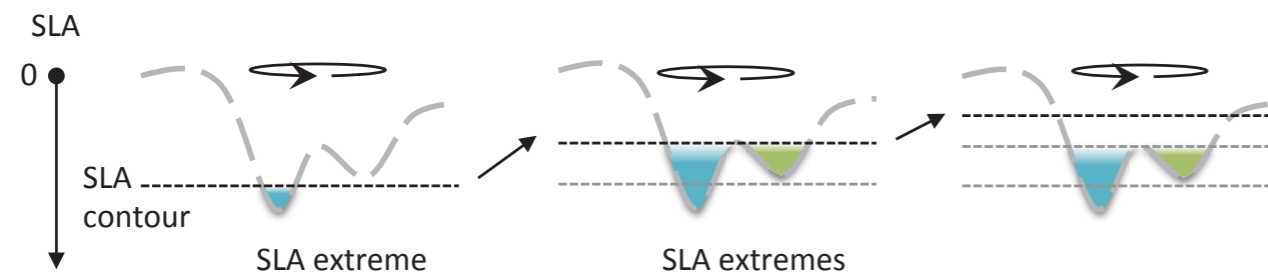
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SLA > 7 cm

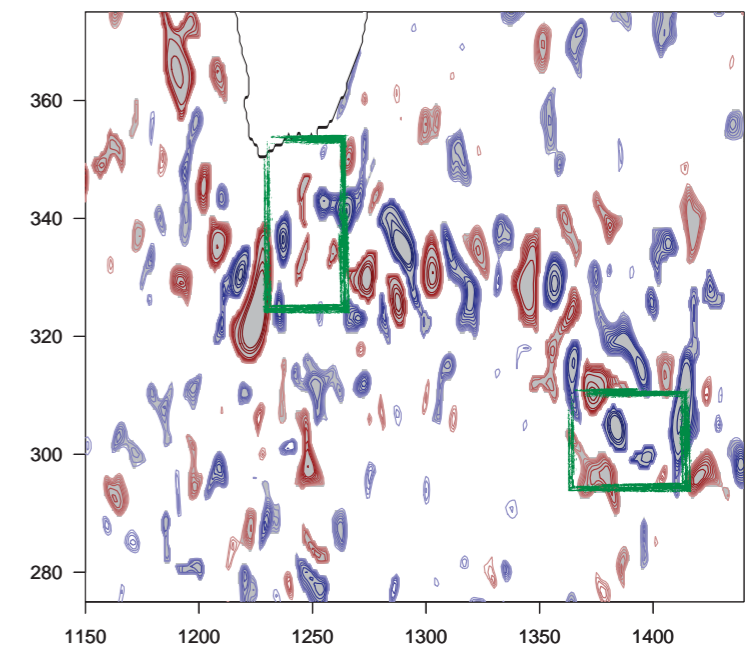


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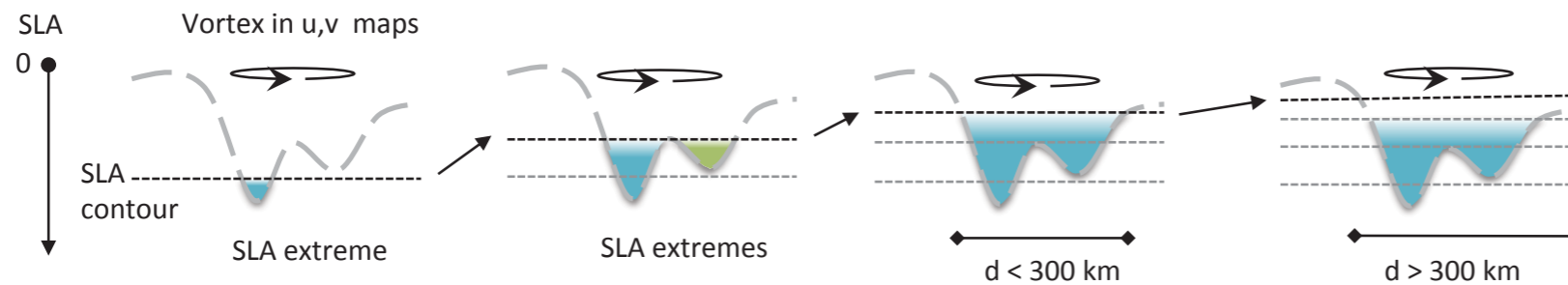


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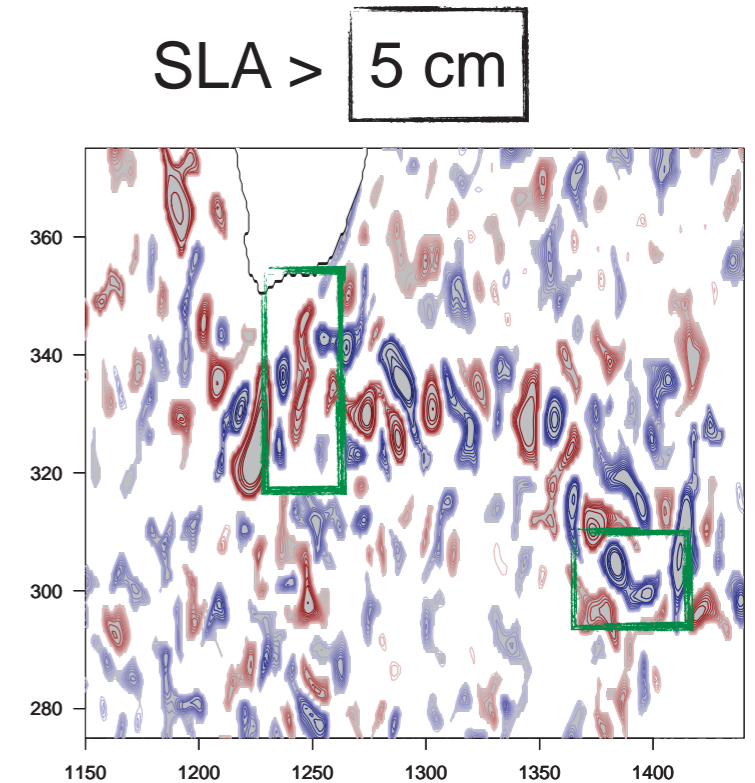
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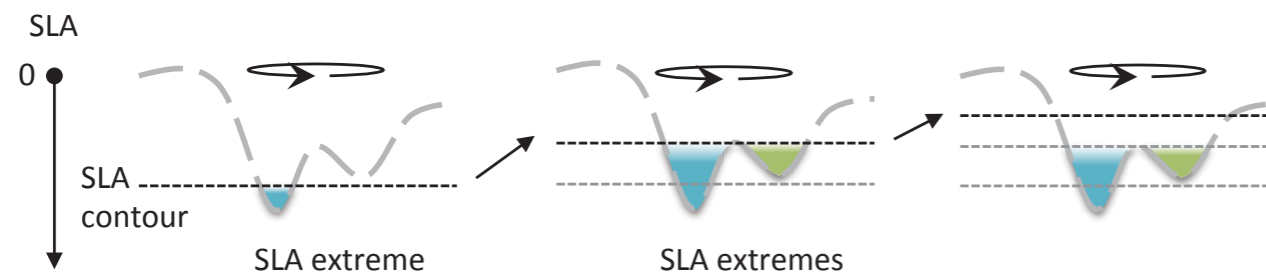


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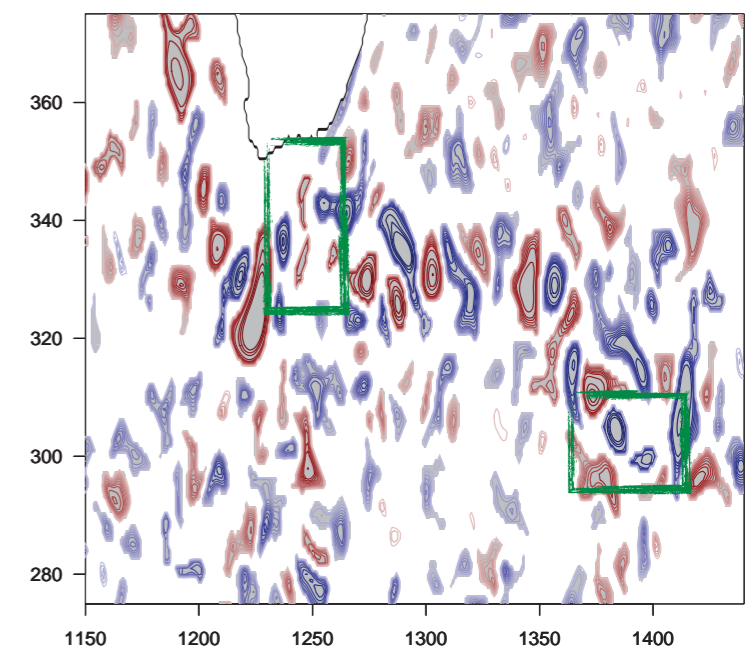


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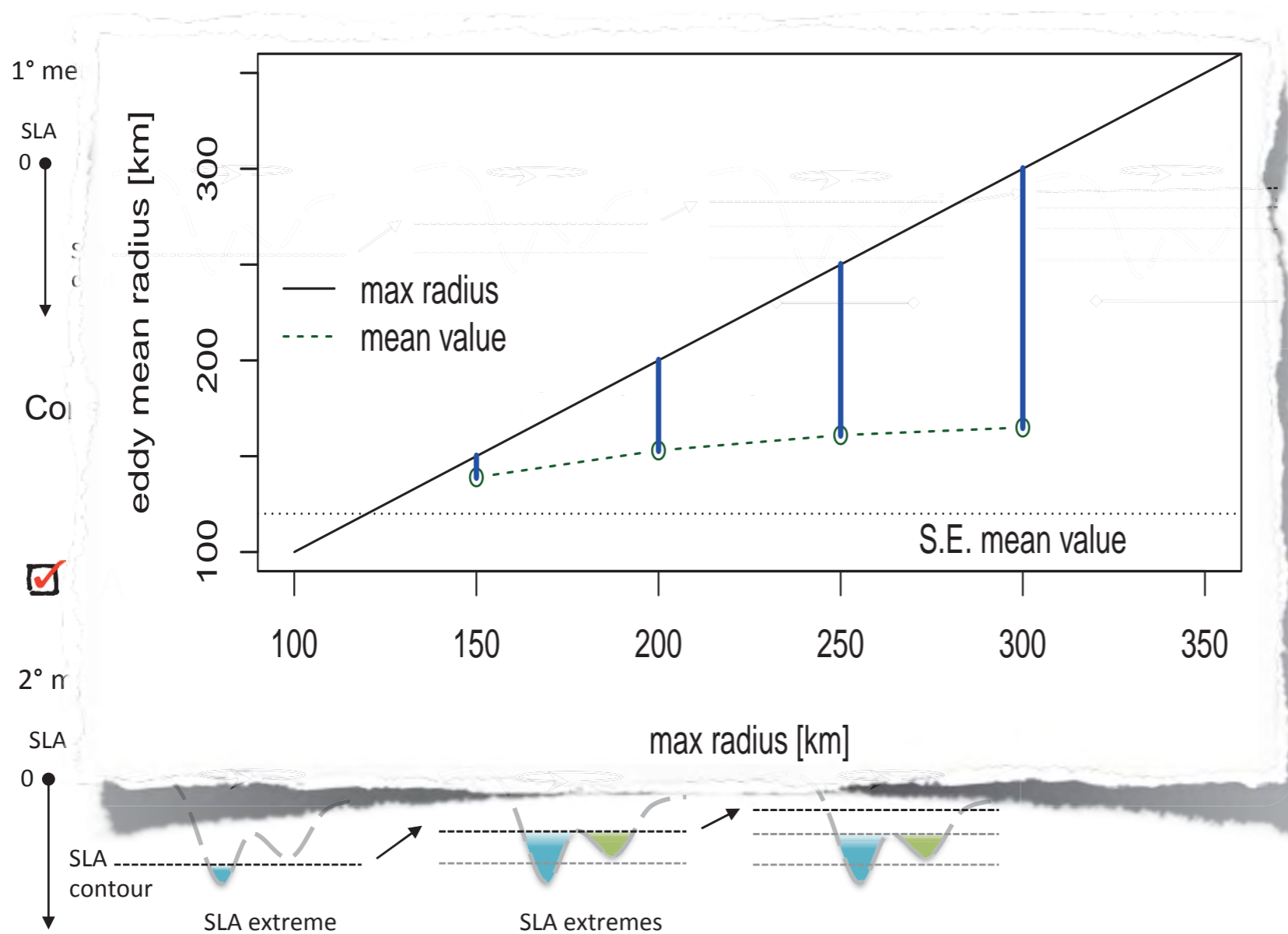
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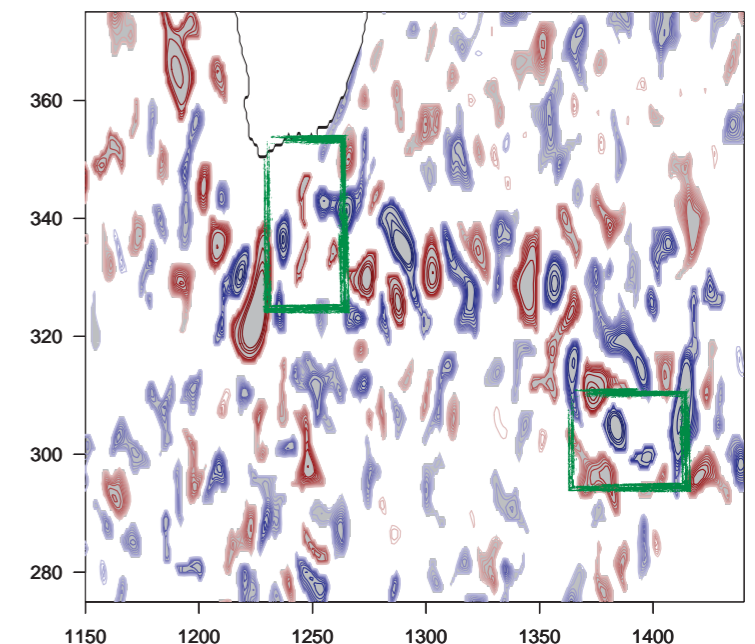
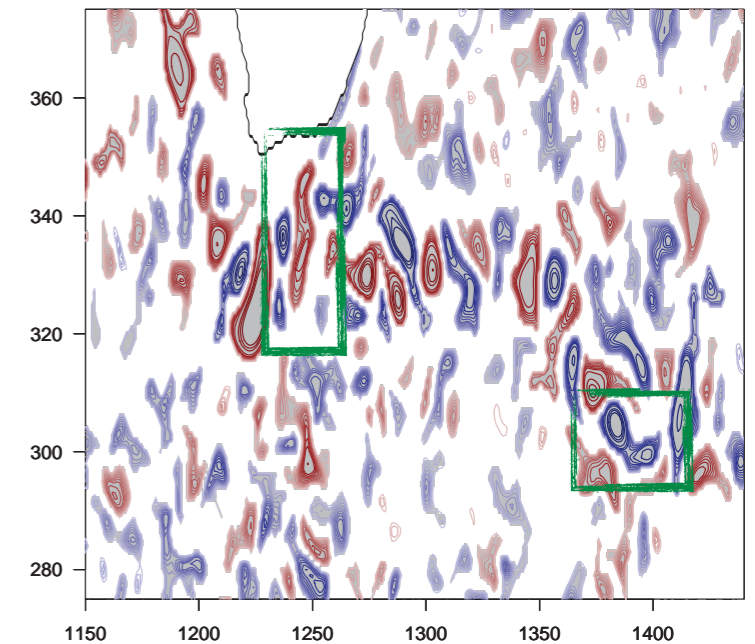
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SLA > 5 cm



Geometry-based method on SLA

Tracking algorithm: nearest-neighbor method

Eddy at time $k+1$ are tracked to

- closest eddy at k
- best area ratio at time k if
 - more than one eddy within 4 points
 - size eddy ratio is 2 (or 1/2) times bigger

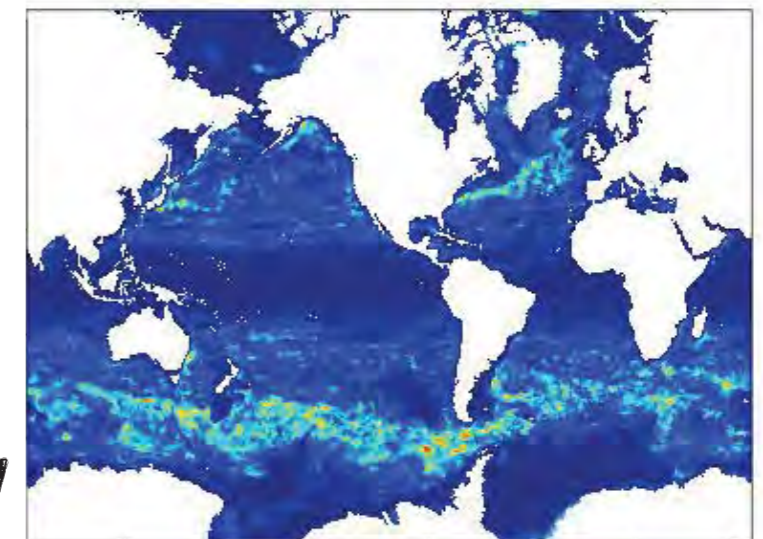
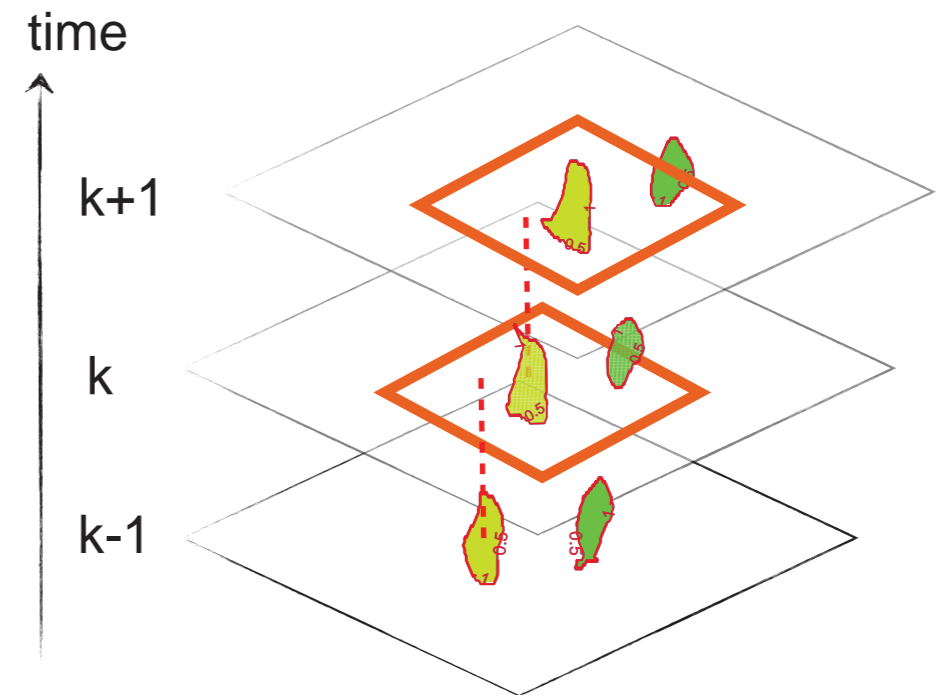
max eddy speed : 25 cm/s

box of 150 km radius for weekly maps

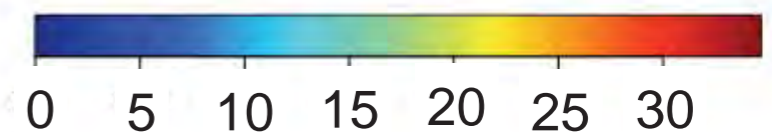
A further tracking between k and $k+2$ is implemented

- Fraction of eddies that are lost and re-found as function of position. Contribution to **long-lived eddies** (> longer than 16 weeks)

minimum eddy size imposed is > 25 km of radius



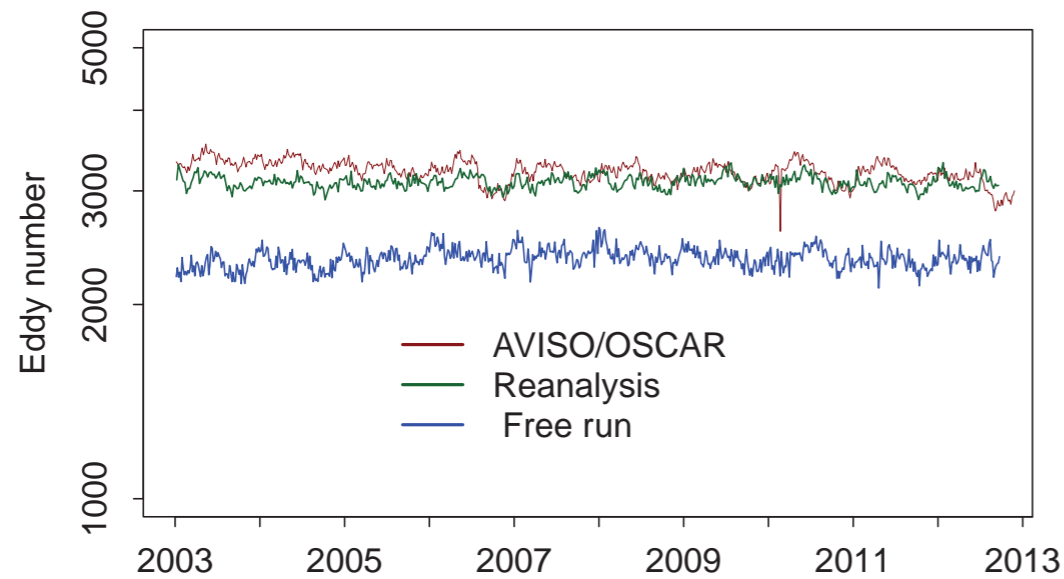
% eddies



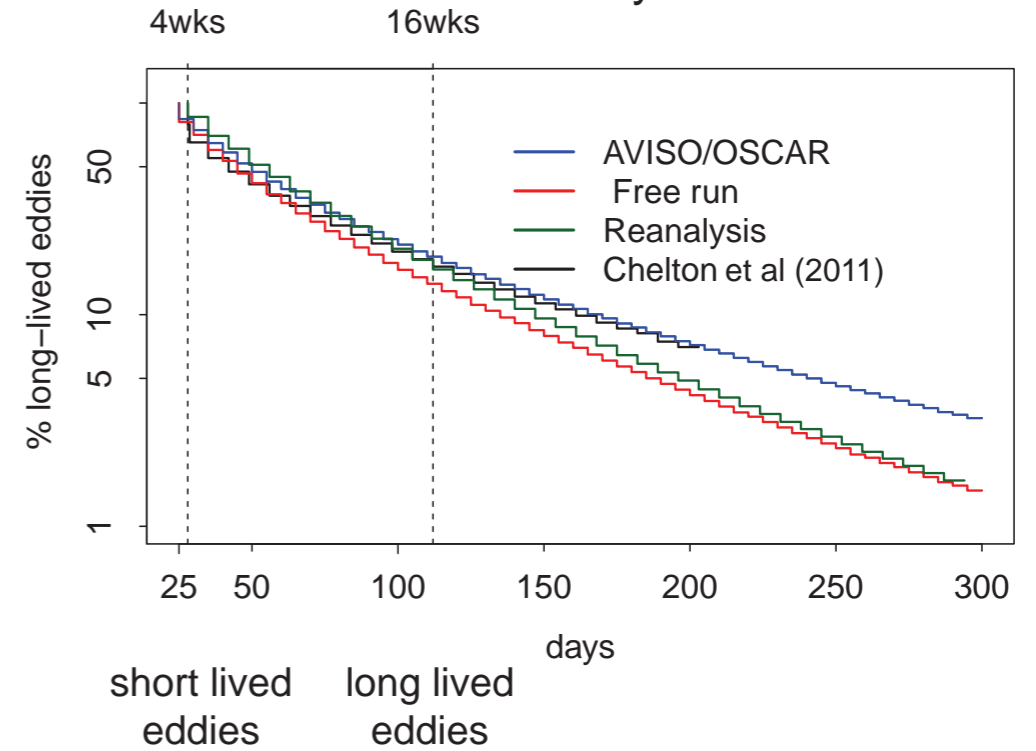
10-year statistics on Global Ocean

10 years statistics on Global ocean basin

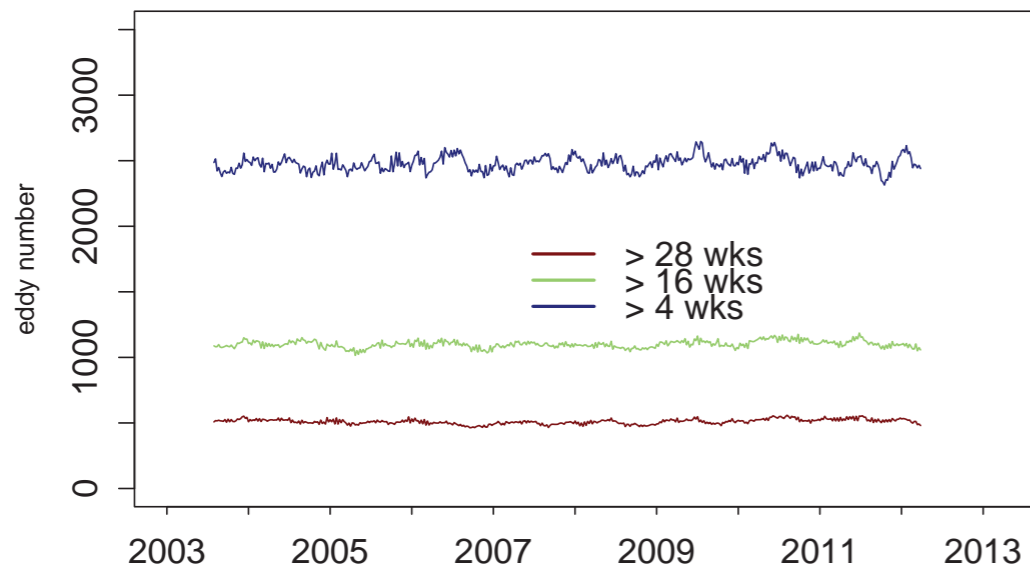
number of eddies per record



decay rate

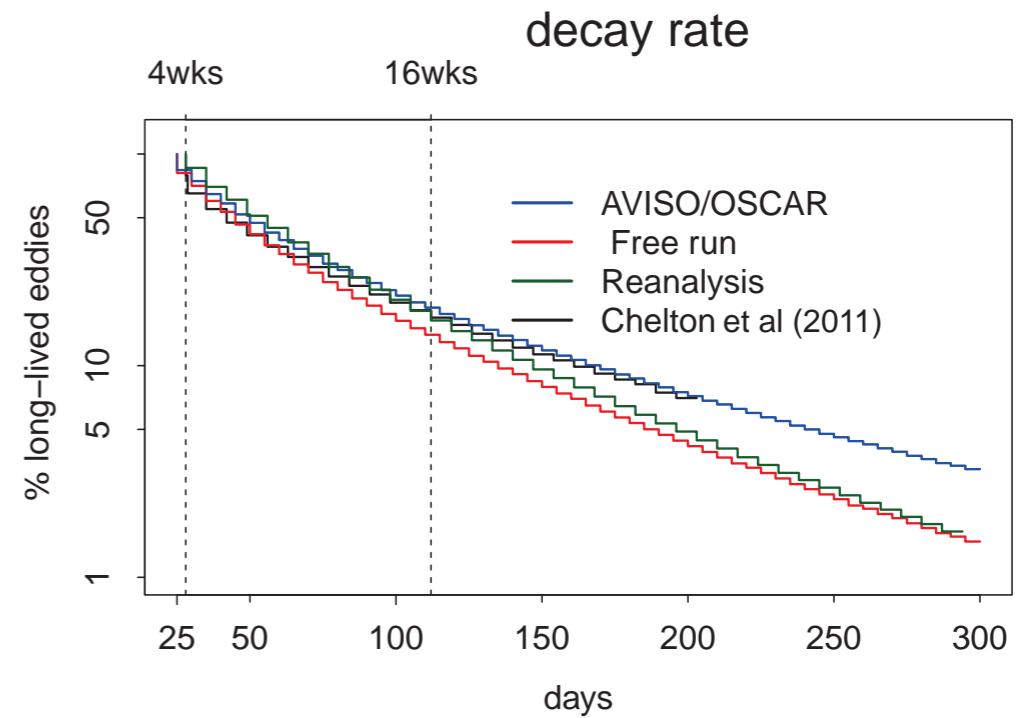
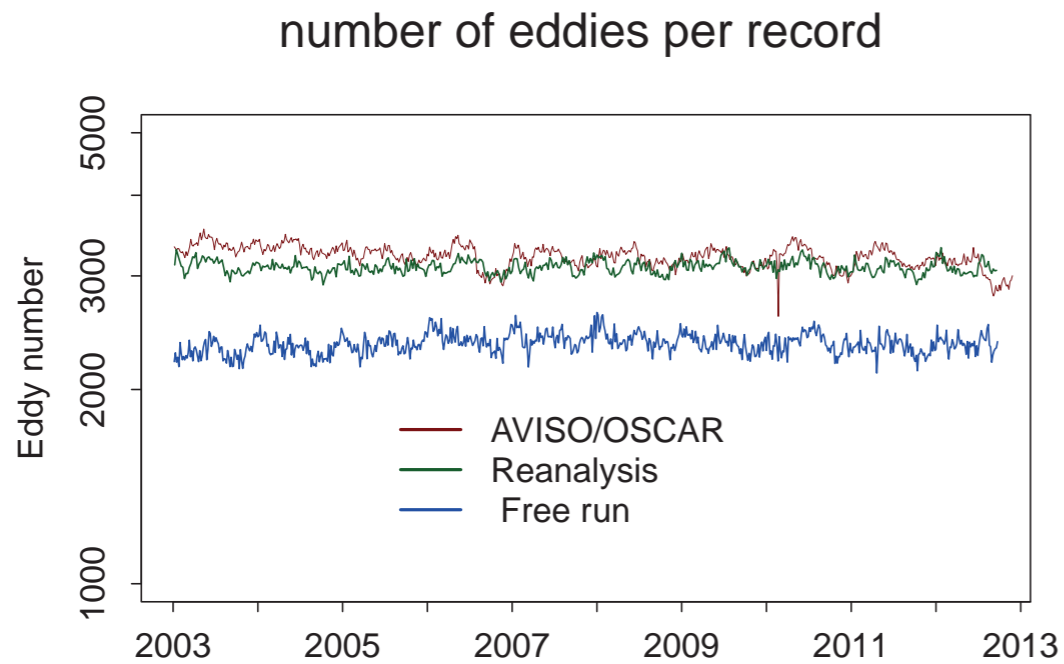


number of eddies with lifetime longer than 4,16,28 weeks

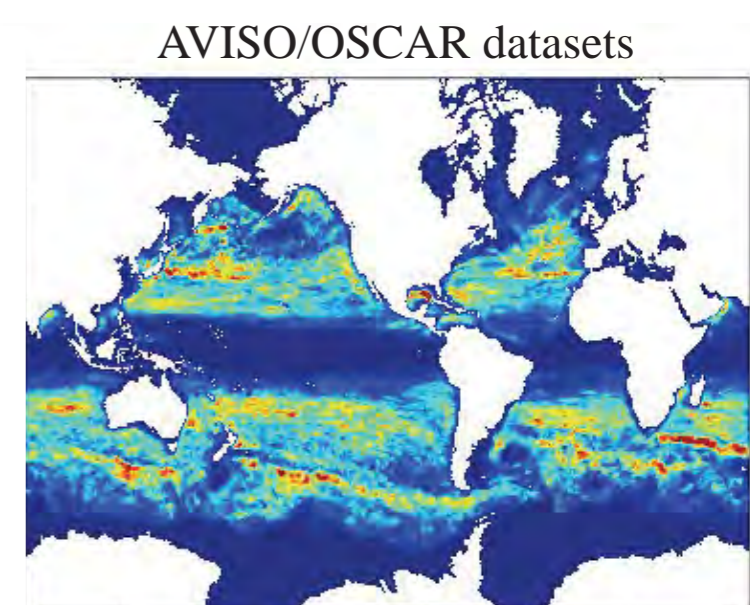
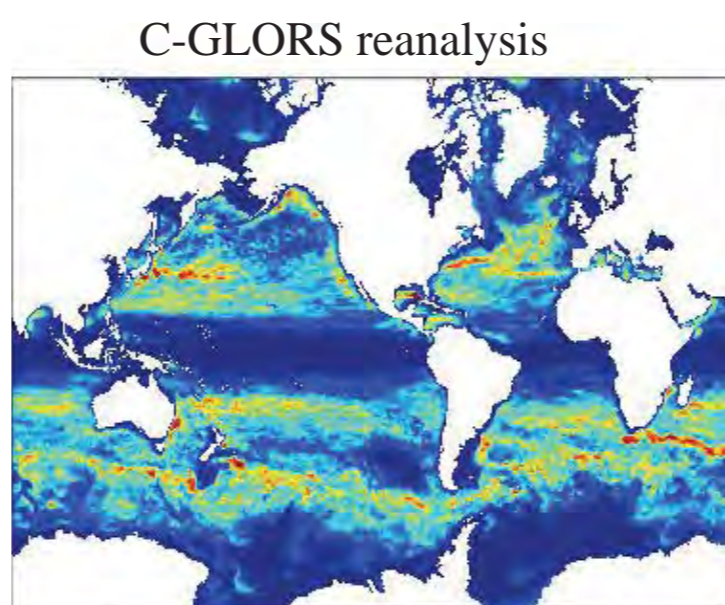
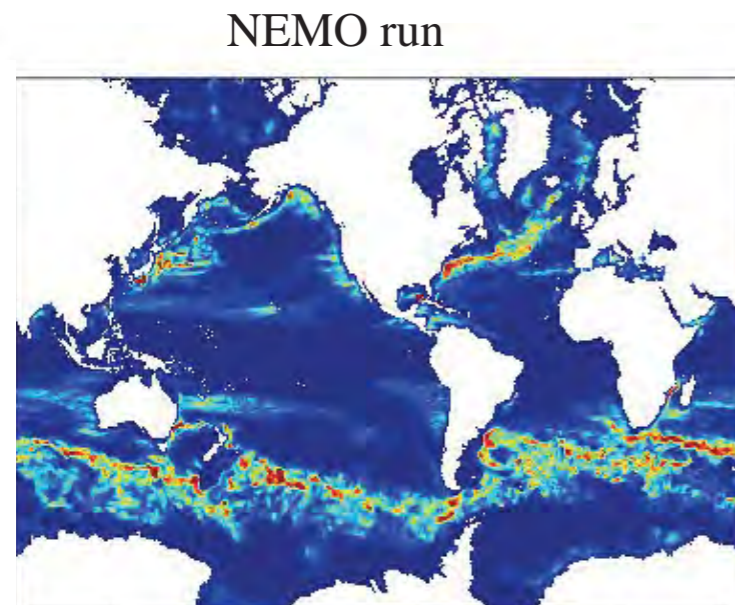


Reanalysis dataset

10 years statistics on Global ocean basin



□ Fraction of eddy occurrence in 10 yrs for eddies with lifetime longer than 16 weeks



% occurrence

% occurrence

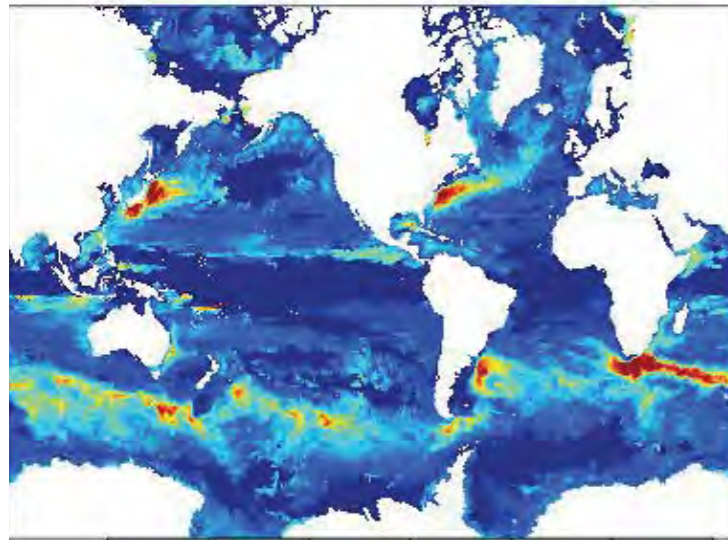
% occurrence



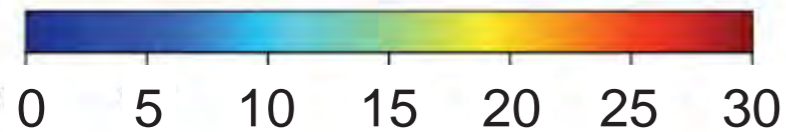
10 years statistics on Global basin

- Mean amplitude for eddies with lifetime longer than 16 weeks

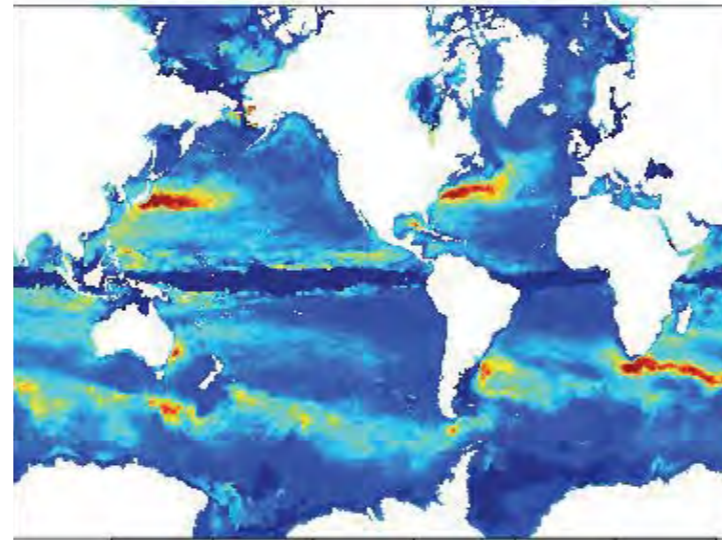
NEMO run



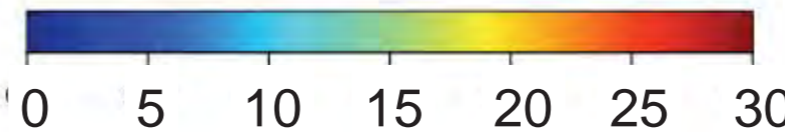
cm



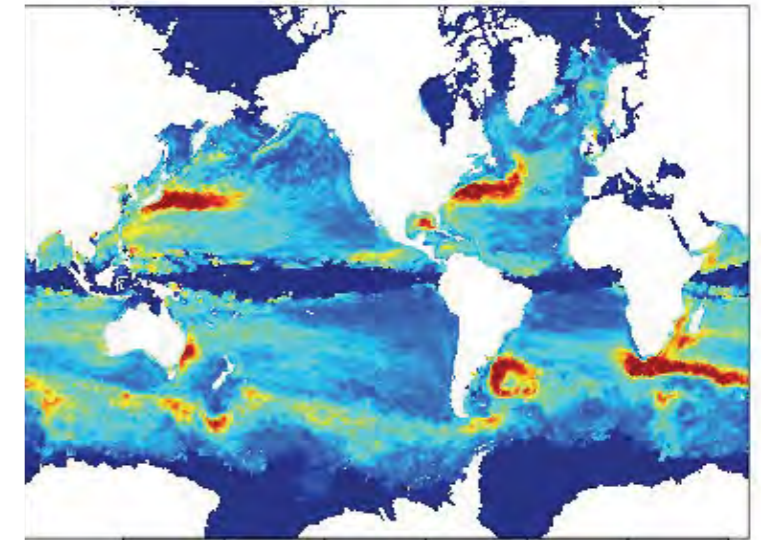
C-GLORS reanalysis



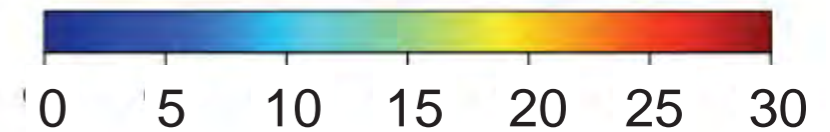
cm



AVISO/OSCAR datasets

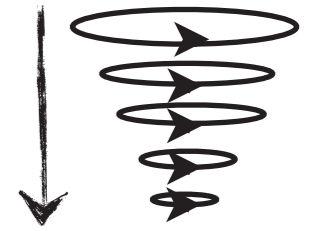


cm



10 years statistics on Global basin

- Trajectories of eddies with lifetime longer than 16 weeks Vs depth

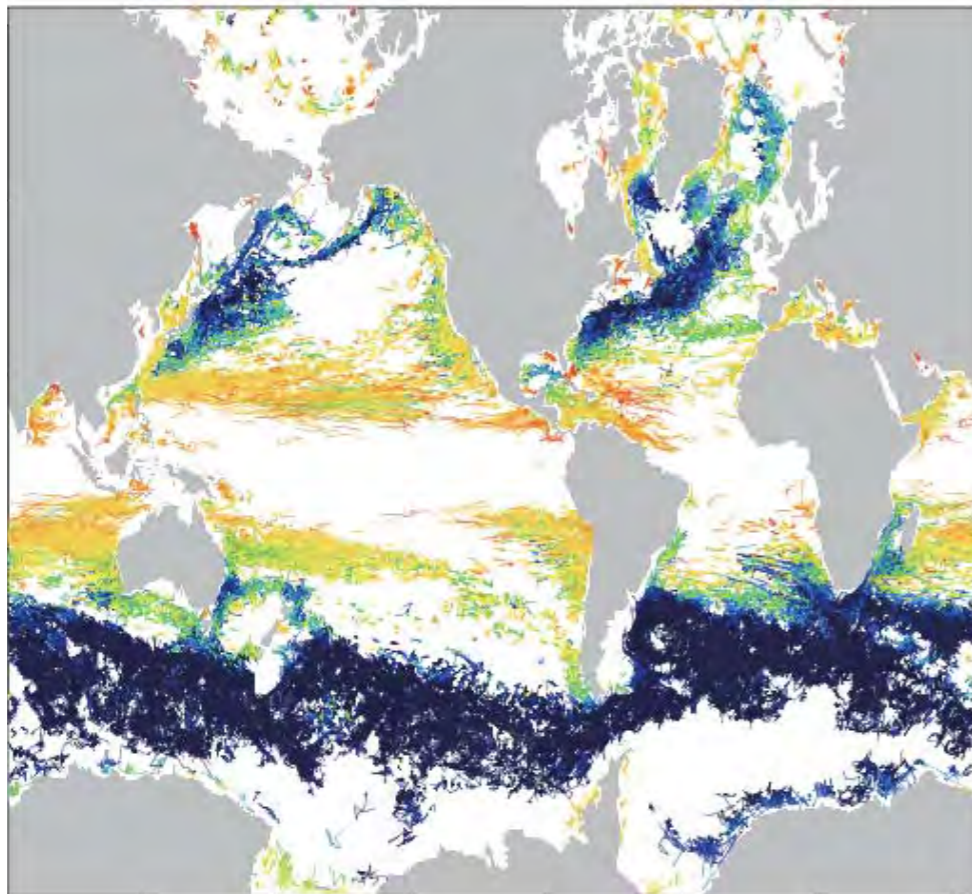


13925 westward

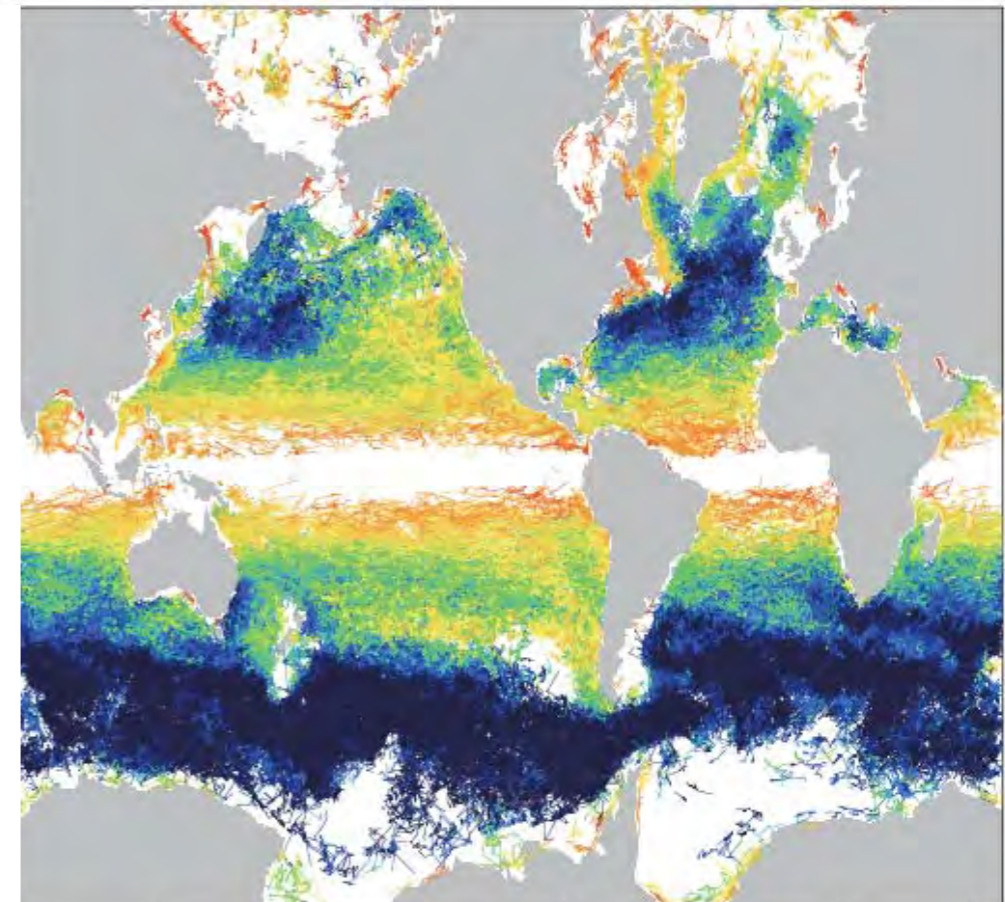
6436 eastward

26961 westward

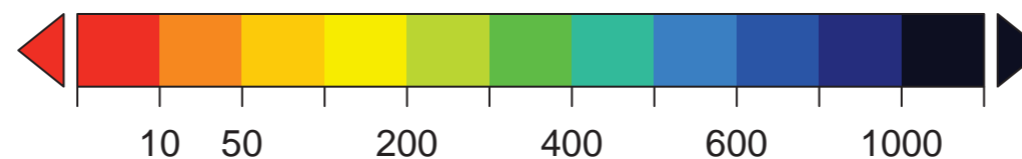
11513 eastward



NEMO run



C-GLORS reanalysis



Mean Depth [m]

- Fraction of total EKE, transported by eddies with lifetime longer than 16 weeks

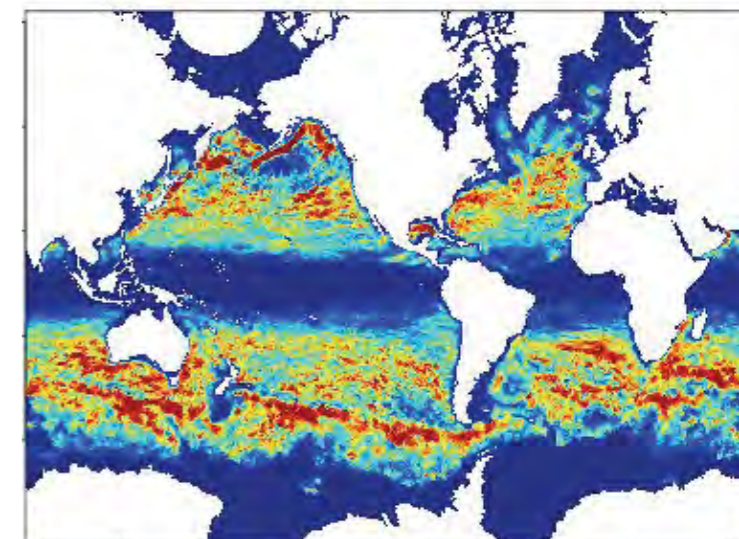
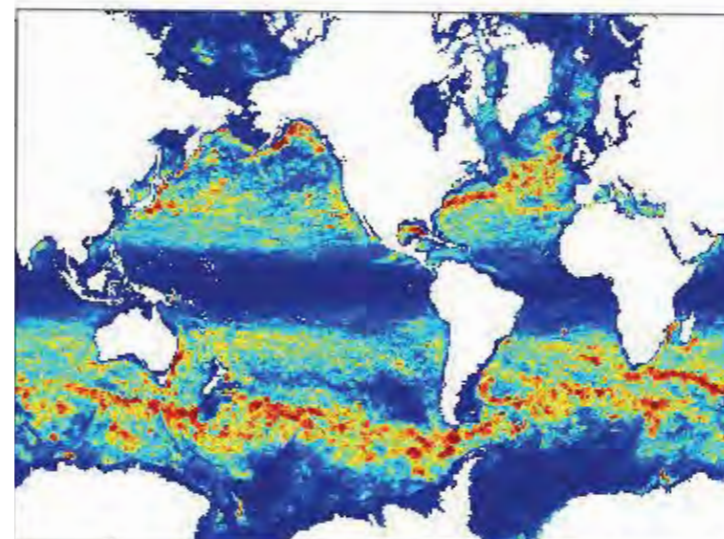
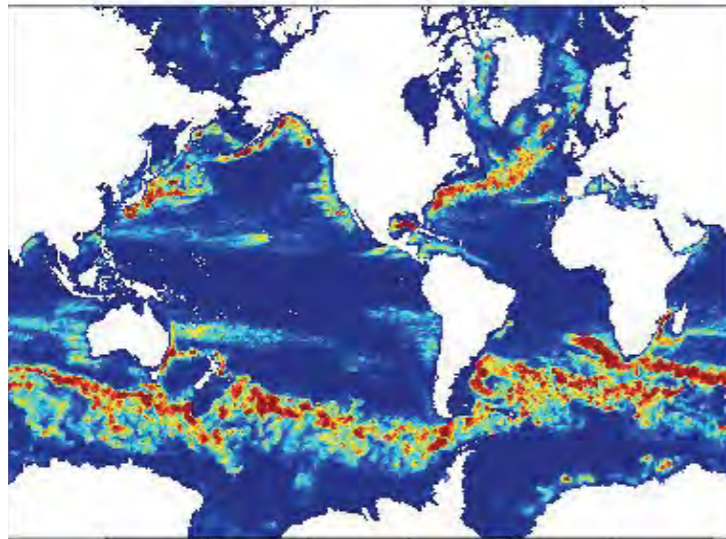
$$\frac{EKE_{\text{eddy}}}{EKE_{\text{tot}}}$$

NEMO run

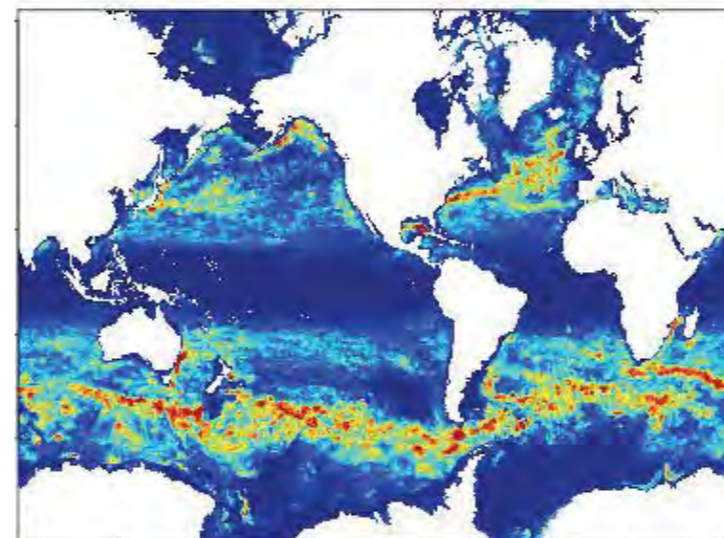
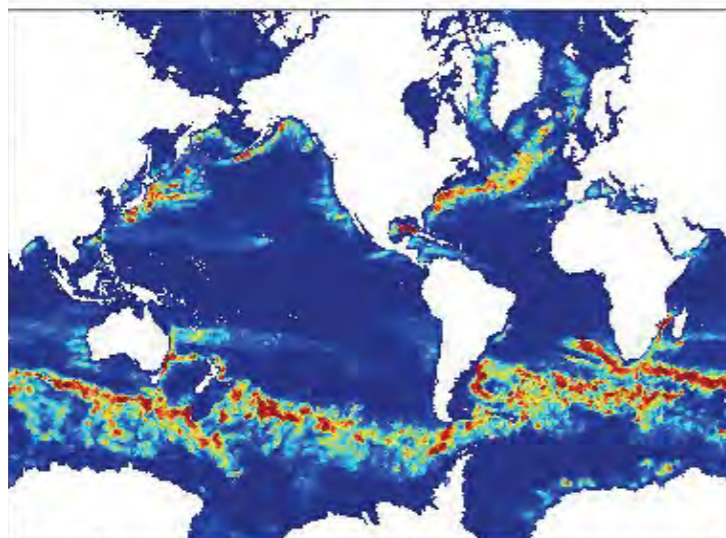
C-GLORS reanalysis

AVISO/OSCAR datasets

Surface



3D

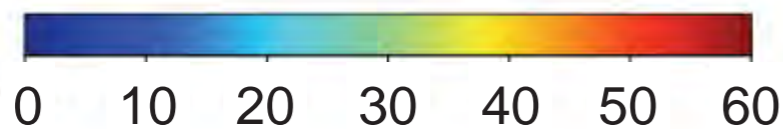


vertical integration stops
when eddy vanish

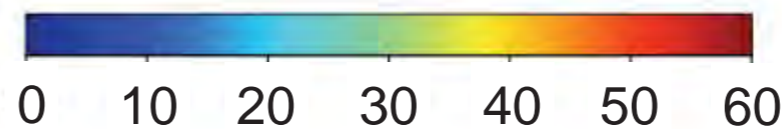
$$\frac{EKE_{\text{eddy}}}{EKE_{\text{tot}}}$$

integrate over the full water
column

% EKE



% EKE



- Fraction of total EKE, transported by eddies with lifetime longer than 4 weeks

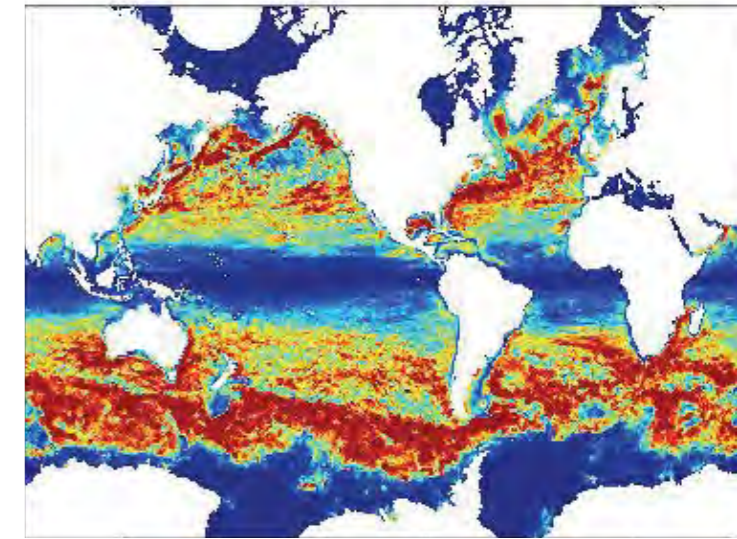
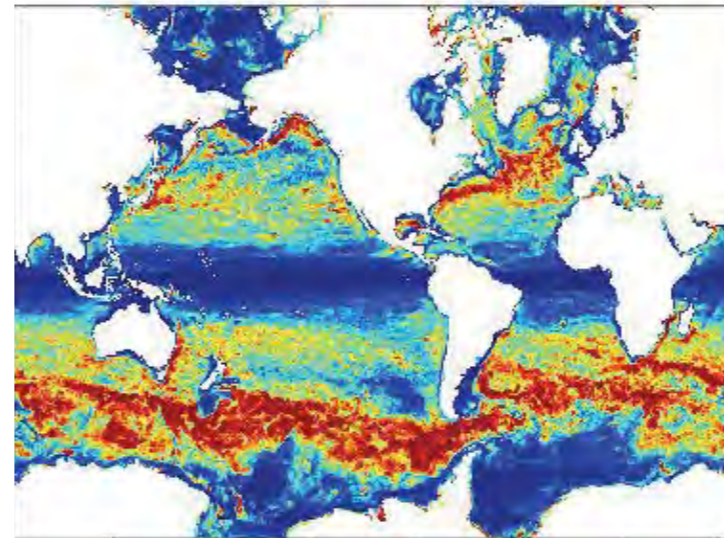
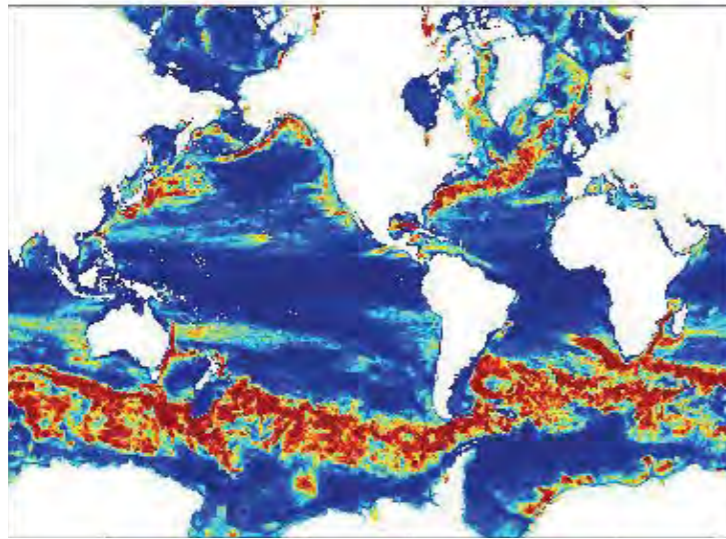
$$\frac{EKE_{\text{eddy}}}{EKE_{\text{tot}}}$$

NEMO run

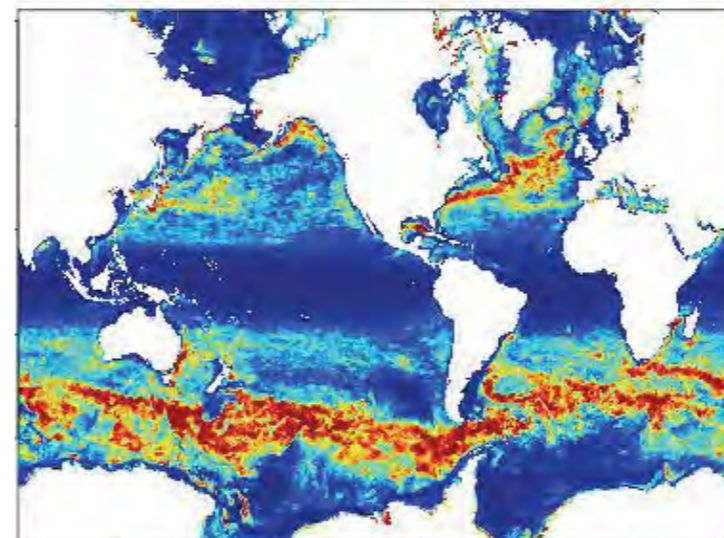
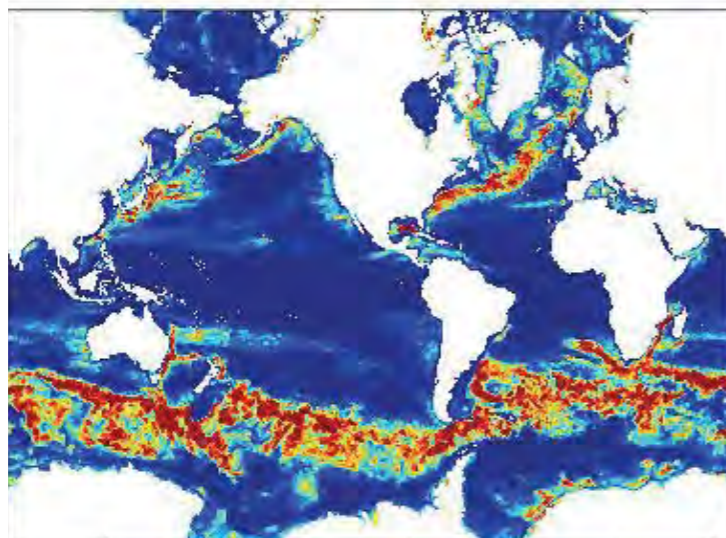
C-GLORS reanalysis

AVISO/OSCAR datasets

Surface



3D



vertical integration stops
when eddy vanish

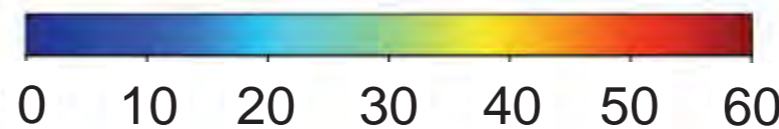
$$\frac{EKE_{\text{eddy}}}{EKE_{\text{tot}}}$$

integrate over the full water
column

% EKE



% EKE



Summary

- Eddies populate any corners of ocean basin, being responsible for diverse phenomenologies

- Eddy census gather three different datasets to highlight the impact of assimilation:
 - 1) NEMO free simulation at 1/4
 - 2) C-GLORS reanalysis at 1/4
 - 3) observed datasets (AVISO/OSCAR) at 1/4

- Assimilation seems to
 - @ global level recover most of the mesoscale variability from satellite/in-situ data and consistently generate variability in the interior

 - @ local level correct the behavior of each single eddy towards a more realistic profiles

- NEXT** Can we give some “trustful” estimates of heat and freshwater fluxes at global level ??

Summary

NEXT How it performs over an $1/16^\circ$ eddying simulation of the global sea ice-ocean system ??



*Iovino, **AC** et al., GMDD 2016*

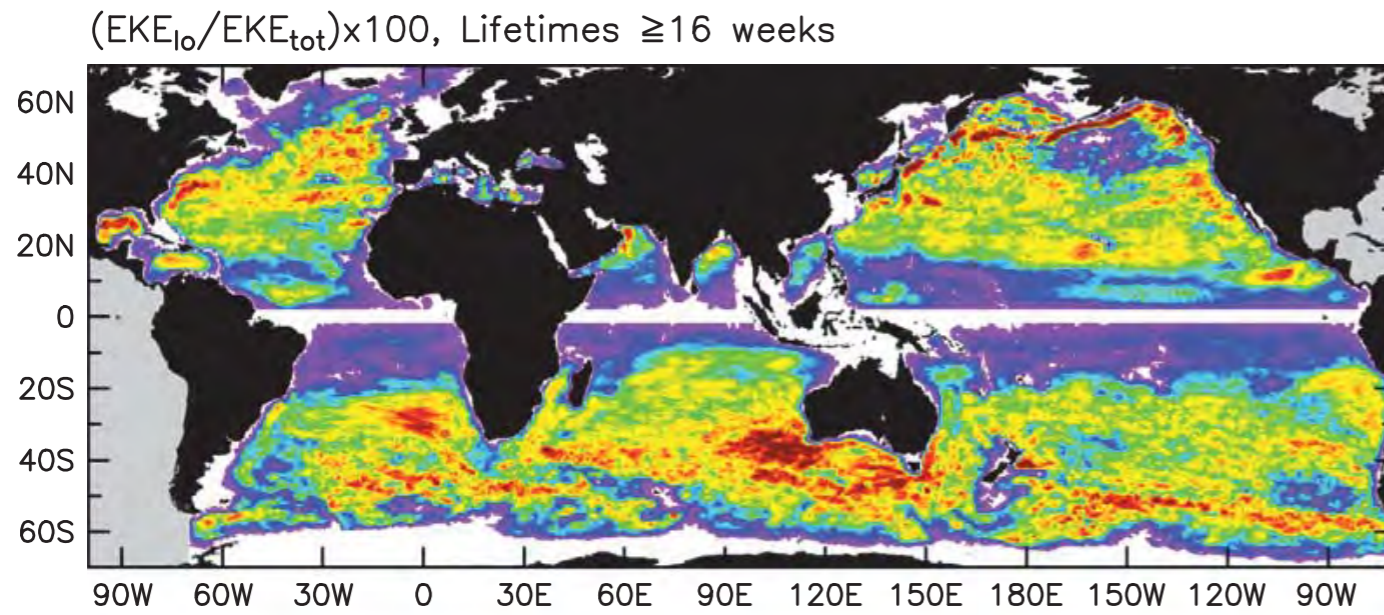
*Stepanov, **AC** et al., JGR:Ocean 2016*

*Stepanov, **AC** et al., JGR:Ocean 2016*

Thank you!

Backup slides

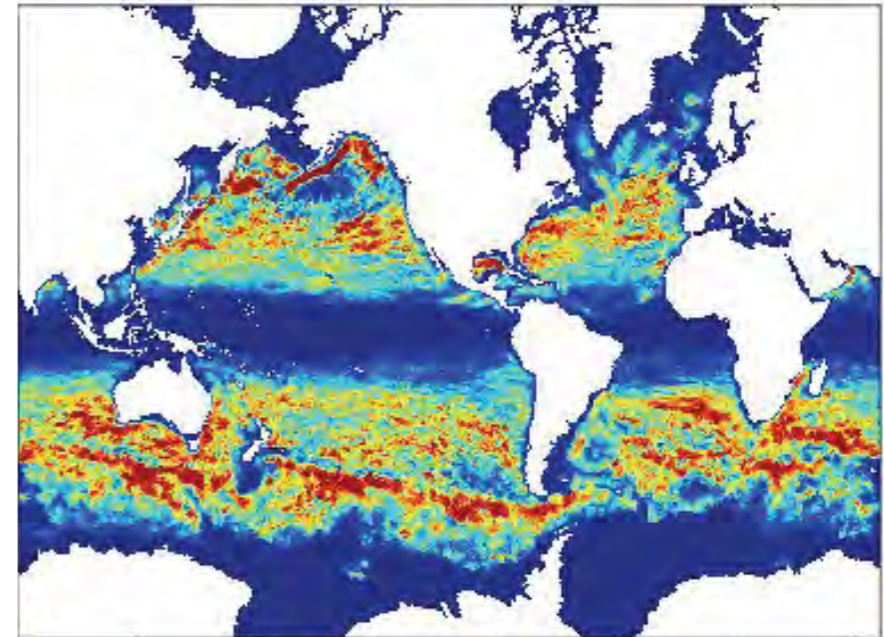
AVISO 1/3



maps averaged over boxes $1^\circ \times 1^\circ$

(Chelton et al. 2011)

AVISO 1/4 and OSCAR datasets



Regional case study: Peru-Chile current system

Peru-Chile currents

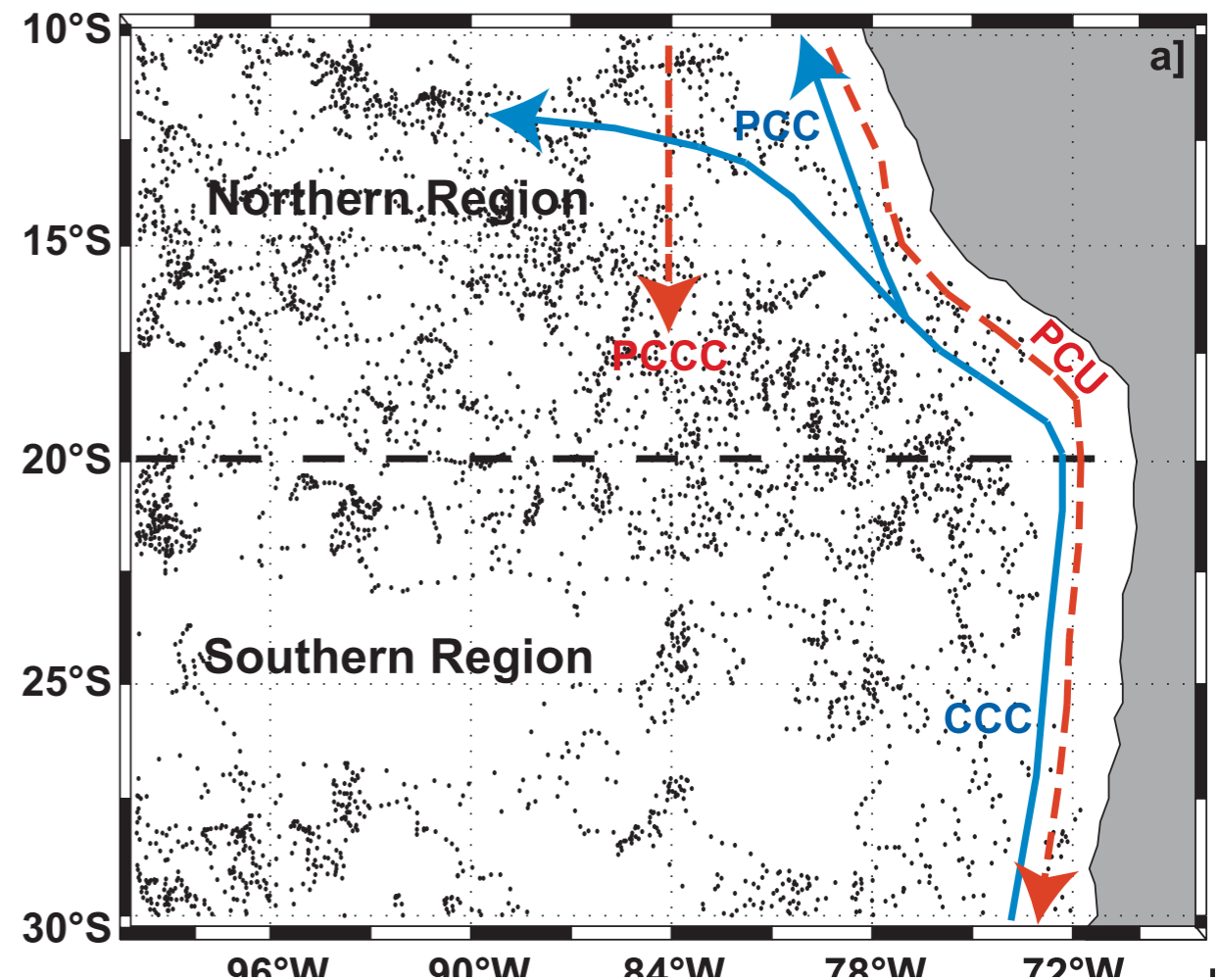
□ The role of eddies on Peru-Chile current system

- ✓ acquire a water mass structure typical of their formation region and propagate offshore
- ✓ In this new environment, eddies appear as anomalous water masses with temperature and salinity anomalies

anti-cyclonic eddies generated from front instabilities of warm poleward currents

cyclonic eddies transport cold water offshore, trapping recently upwelled water

warm, salty sub-surface subtropical water from 9° S in fall (austral spring)



(Chaigneau et al. 2011)

near-coastal cold upwelling region

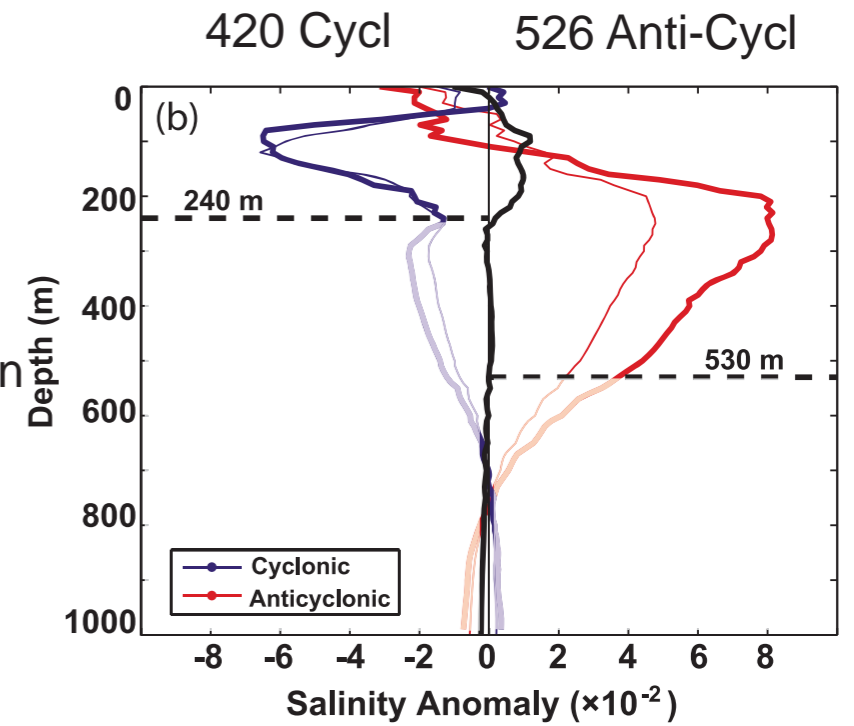
Regional case study

□ Salinity anomaly trapped in the eddies

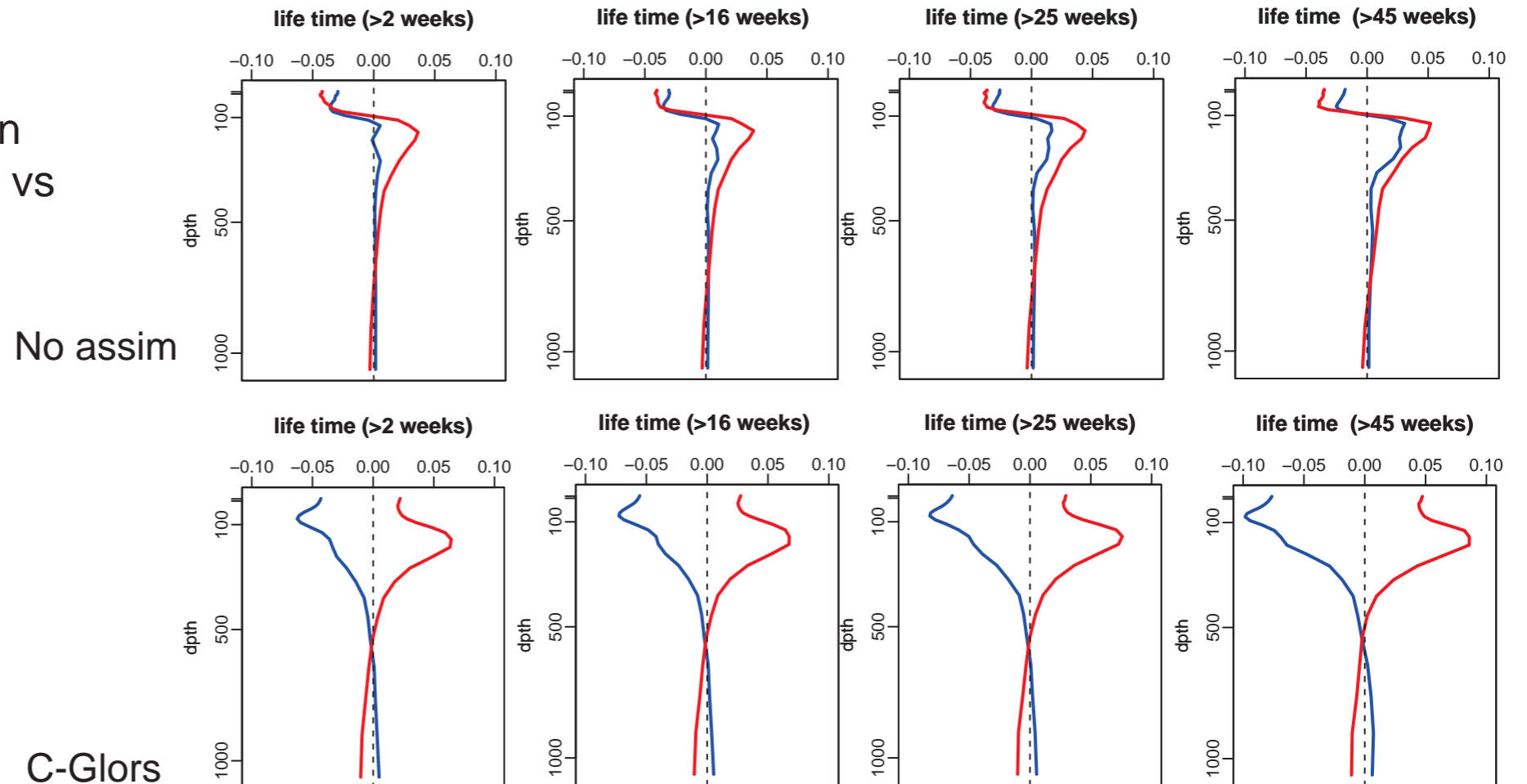
AE (red)

CE (blue)

Argo profile Dec 2003-Oct 2009.
Anomaly eval. vs climatological mean
from CSIRO Atlas of Regional Seas
(Chaigneau et al. 2011)



☑ Reanalysis corrects mean amount as well as profile in fall season. Anomaly eval. vs climatological mean from 10ys run



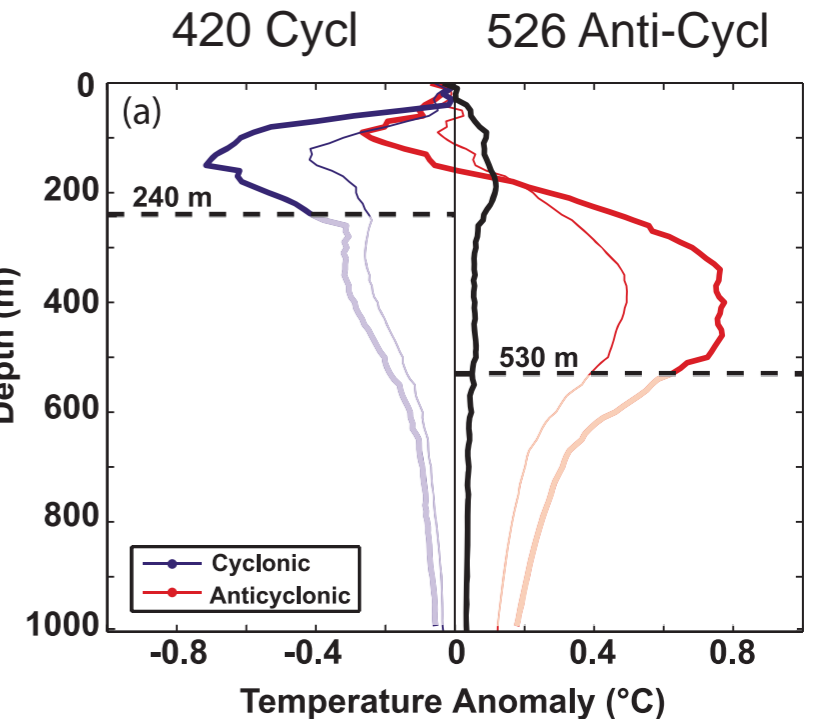
Regional case study

☐ Temperature anomaly trapped in the eddies

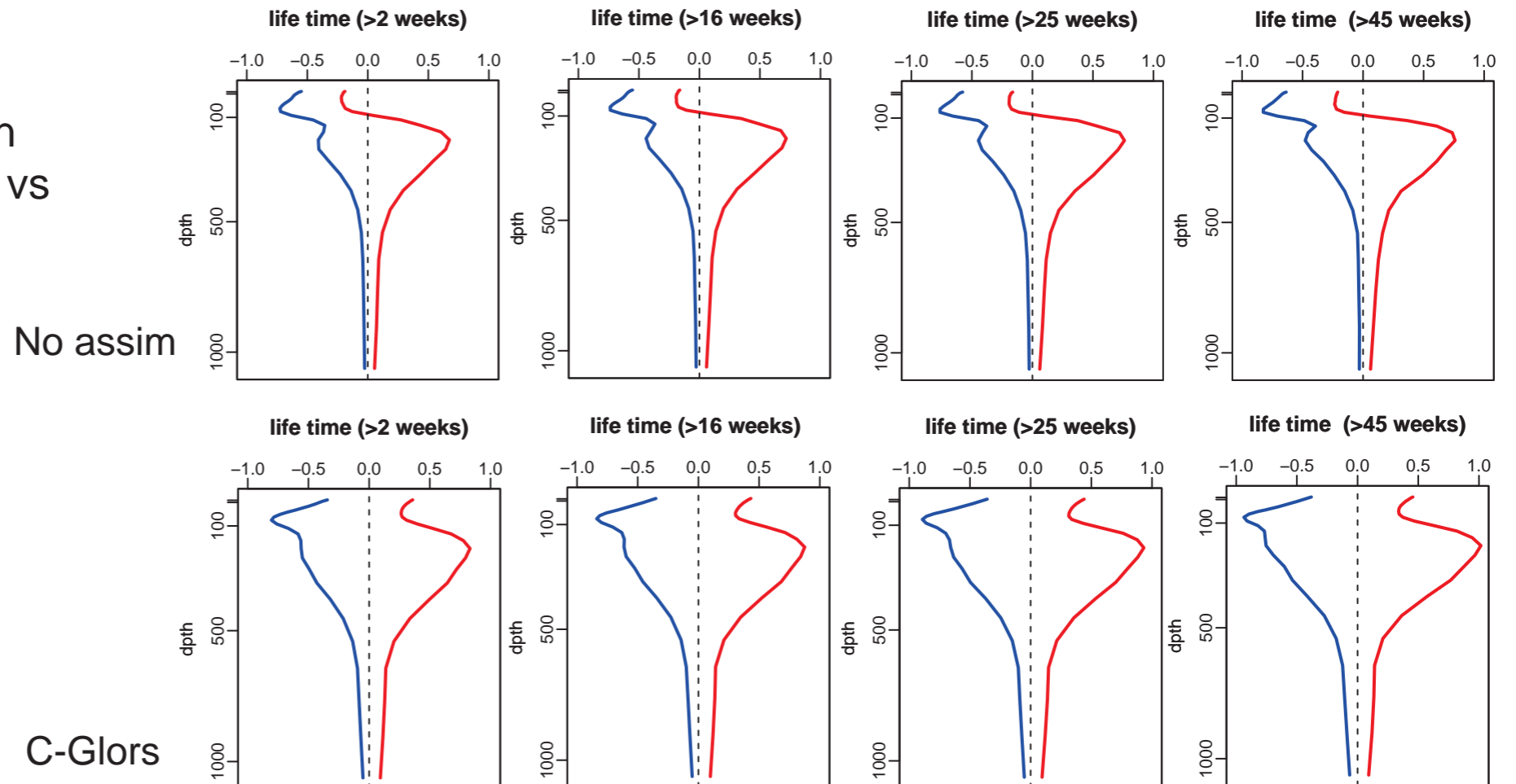
AE (red)

CE (blue)

Argo profile Dec 2003-Oct 2009.
Anomaly eval. vs climatological mean
from CSIRO Atlas of Regional Seas
(Chaigneau et al. 2011)



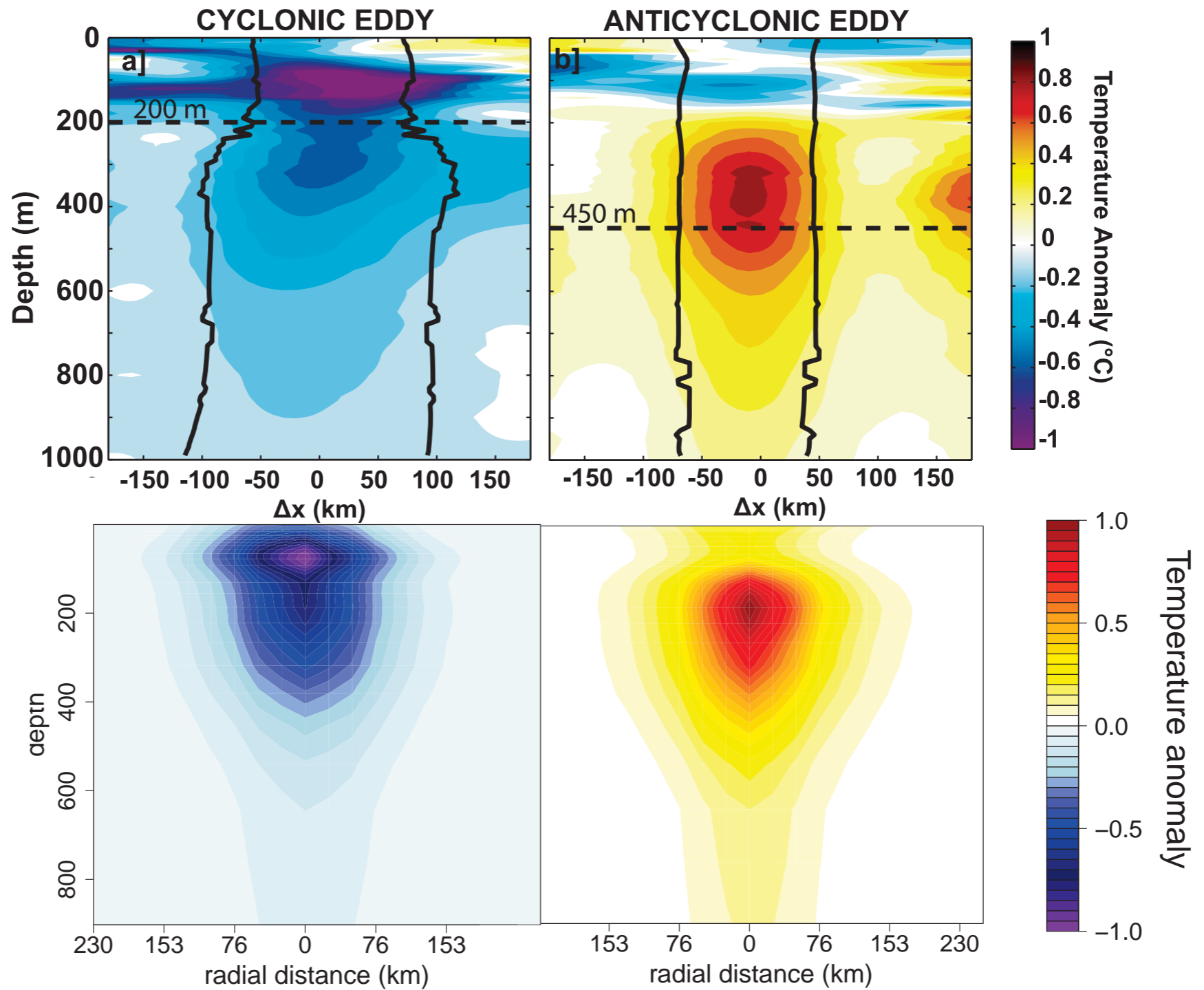
☑ Reanalysis corrects mean amount as well as profile in fall season. Anomaly eval. vs climatological mean from 10ys run



Regional case study

□ Section of eddy

(Chaigneau et al. 2011)



C-GLORS