

# Lagrangian geographies of the Gulf of Mexico

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# Objectives

Using trajectories of floats and drifters in the Gulf of Mexico (GoM):

- Subdivide the GoM and the deep GoM into regions with similar dynamics
- Identify almost invariant regions and their respective timescale
- Assess connectivity
- Forecast density evolutions (oil spill, plastic, larvae, etc)

# Data sets for surface and deep GoM

## Drifters

- 3500+ drifter trajectories from several different sources 1992–2018 (GDP<sup>1</sup>, CARTHE<sup>2</sup>, CICESE, SCULP, AOML<sup>2</sup>, USCG<sup>2</sup>)

RAFOS experiment sponsored by BOEM (July 2011–May 2015)<sup>3</sup>:

- 121 floats at 1500 m
- 6 profiling floats with RAFOS technology at 1500m
- 31 floats at 2500 m

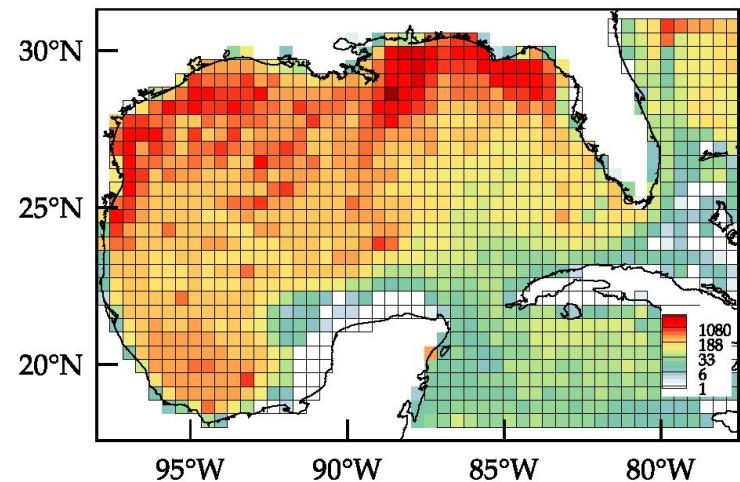
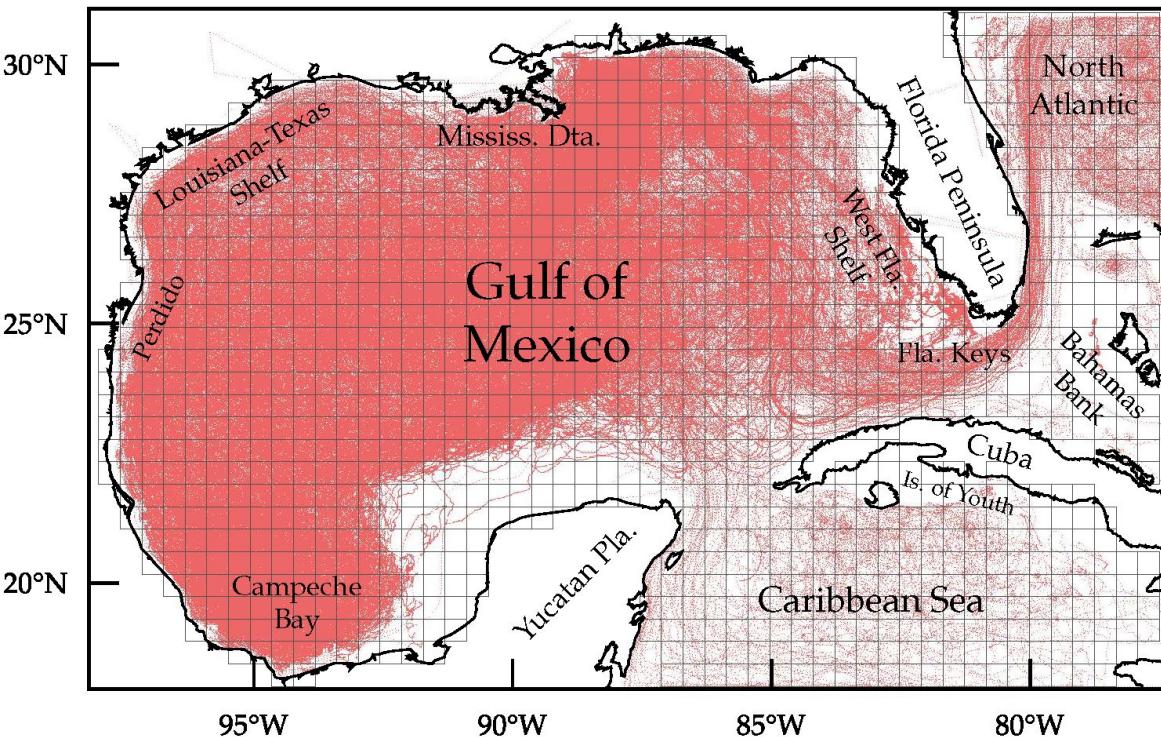
## *Data publicly available*

<sup>1</sup> Physical Oceanography Division Global drifters (PhOD GDP)

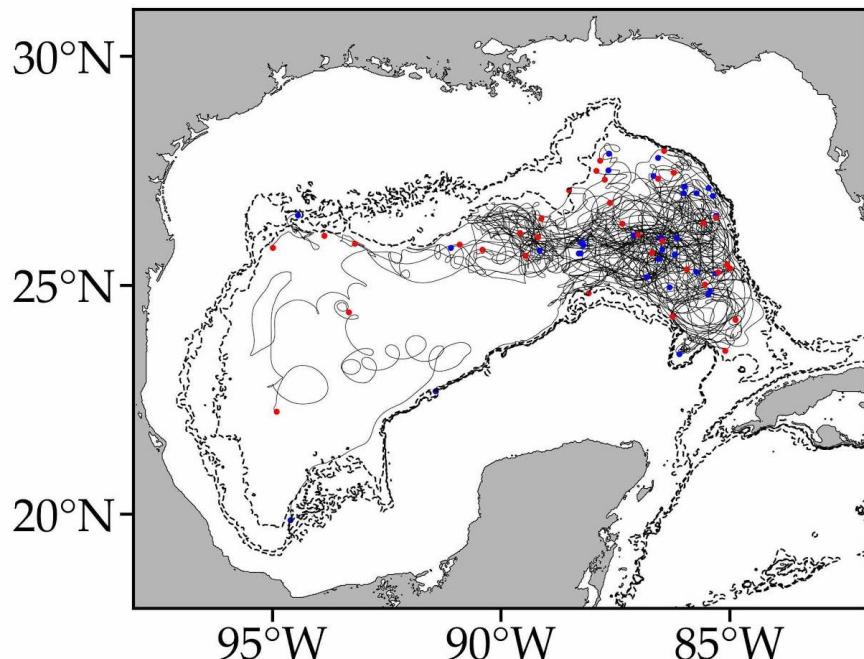
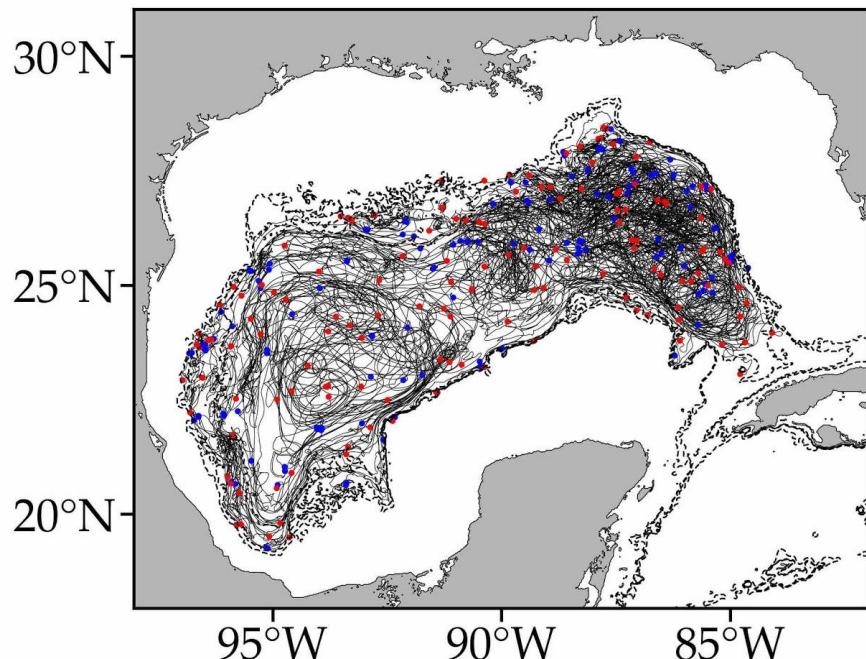
<sup>2</sup> Gulf of Mexico Research Initiative Information & Data Cooperative (GRIIDC)

<sup>3</sup> WOCE Subsurface Float Data Assembly Center (WFDAC)

# Drifter trajectories and density

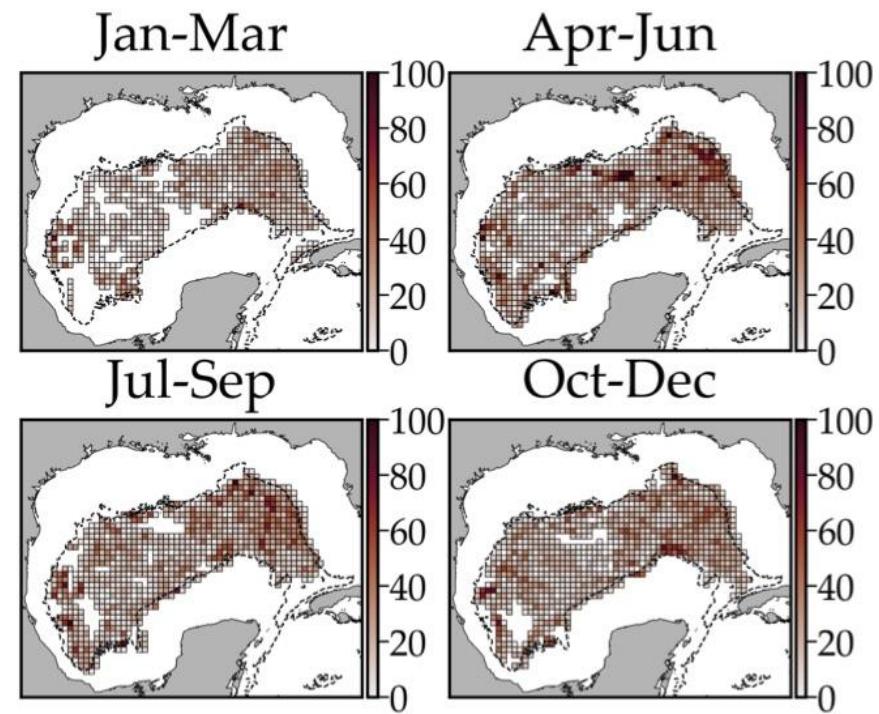
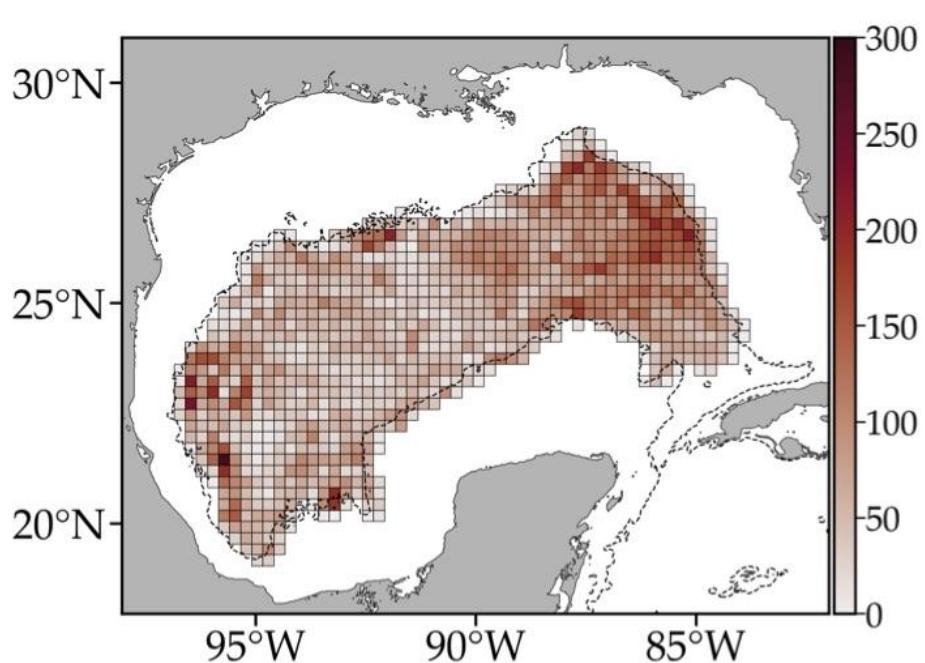


# RAFOS float trajectories



1500 m (left) and 2500 m (right)

# RAFOS float density



Complete data set (left) and seasonal data sets (right)

# Transfer Operator

- Focus on the evolution of probability densities rather than individual trajectories
- transition probabilities are described by a stochastic kernel  $K(x,y)$
- $f(x)$  is a probability density
  - Larger or equal to 0
  - Integral on the domain is 1
- Evolution of the density forward in time

$$\mathcal{P}f(y) = \int_X K(x, y)f(x) \, d x$$

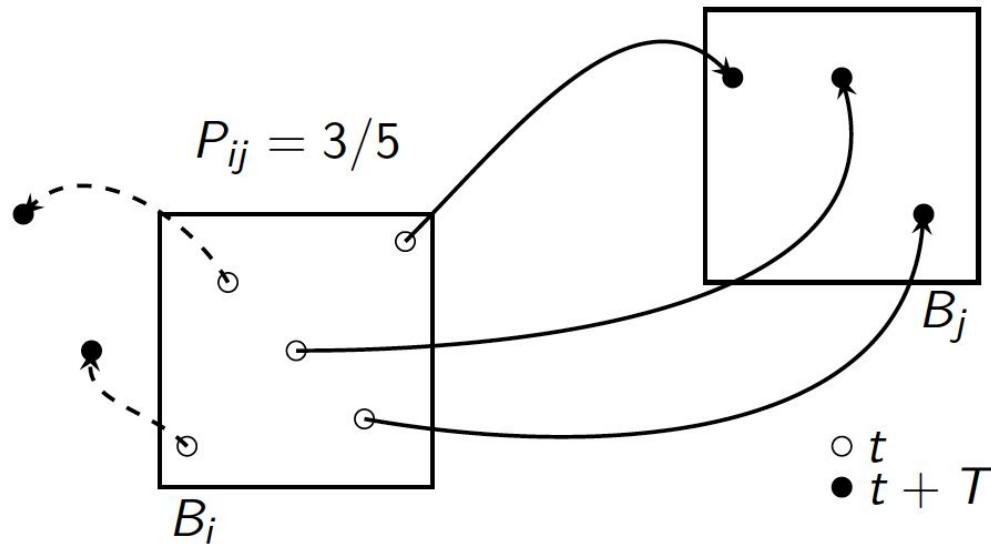
# Discrete version of the Transfer Operator

- Cover the domain with  $N$  connected boxes ( $B_1, B_2, \dots, B_N$ )
- Projected functions in  $L^1(X)$  onto a finite-dimension space  $V_N$
- Perron-Frobenius theorem
- Probability of moving from bin i to j:

$$P_{ij} = \text{prob}[\xi_{t+T} \in B_j \mid \xi_t \in B_i] = \frac{\int_{B_j} \int_{B_i} K(x, y) \, d x \, d y}{\text{area}(B_i)}$$

# Transition matrix

$$P_{ij} \approx \frac{\# \text{ points in } B_i \text{ at } t \text{ that evolve to } B_j \text{ at } t + T}{\# \text{ points in } B_i \text{ at } t}$$



# Evolve a density (pushforward)

- Left multiplication of the discrete density  $f(x)$  with the transition matrix
- Each time evolve the density T days
- If  $P$  is row-stochastic, the mass is conserved and  $\sum f_i = 1$ .

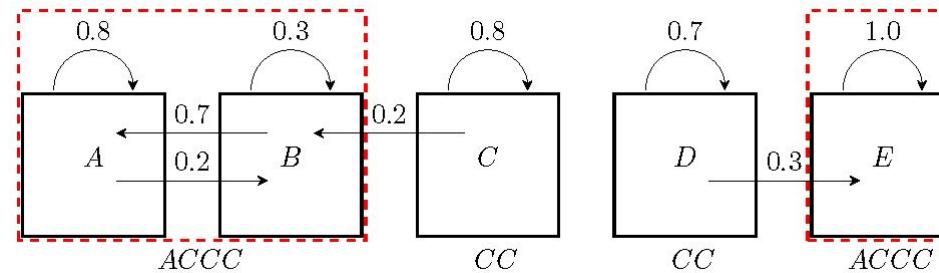
$$\mathbf{f}^{(k)} = \mathbf{f}P^k, k = 1, 2, \dots$$

# Eigenspectrum

Regions where trajectories converge and their basins of attraction are encoded in the eigenvectors of the transition matrix (Froyland, Stuart, and van Sebille 2014).

- Right eigenvector of  $\mathbf{P}$  is the basin of attraction (constraints connectivity)
- Left eigenvector of  $\mathbf{P}$  is the attractor or almost-invariant region

# Example with a simple 5 states problem

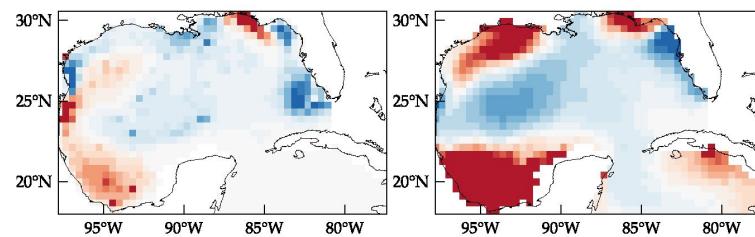
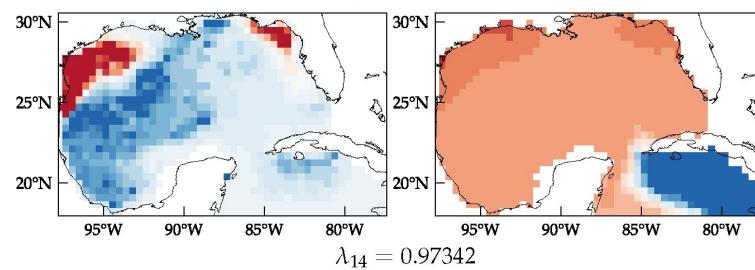
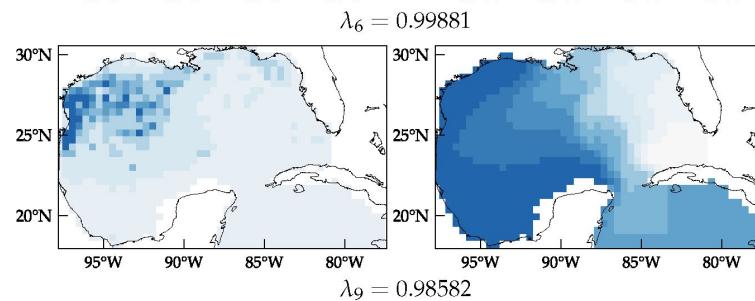
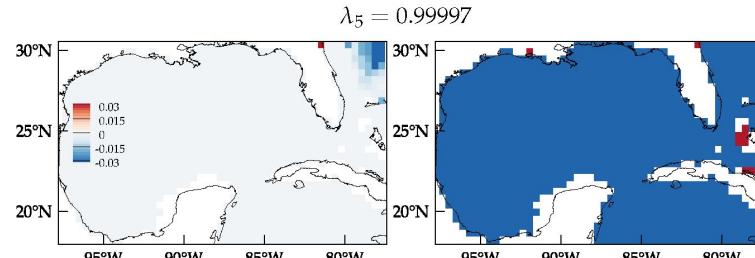

$$\mathcal{P} = \begin{array}{c|ccccc} & A & B & C & D & E \\ \hline A & 0.8 & 0.2 & 0 & 0 & 0 \\ B & 0.7 & 0.3 & 0 & 0 & 0 \\ C & 0 & 0.2 & 0.8 & 0 & 0 \\ D & 0 & 0 & 0 & 0.7 & 0.3 \\ E & 0 & 0 & 0 & 0 & 1.0 \end{array}$$

# Example with a simple 5 states problem

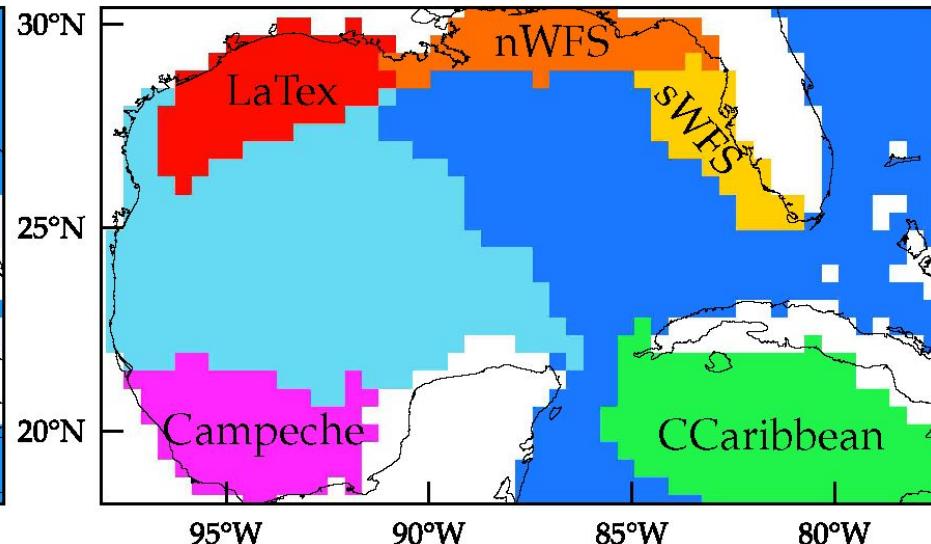
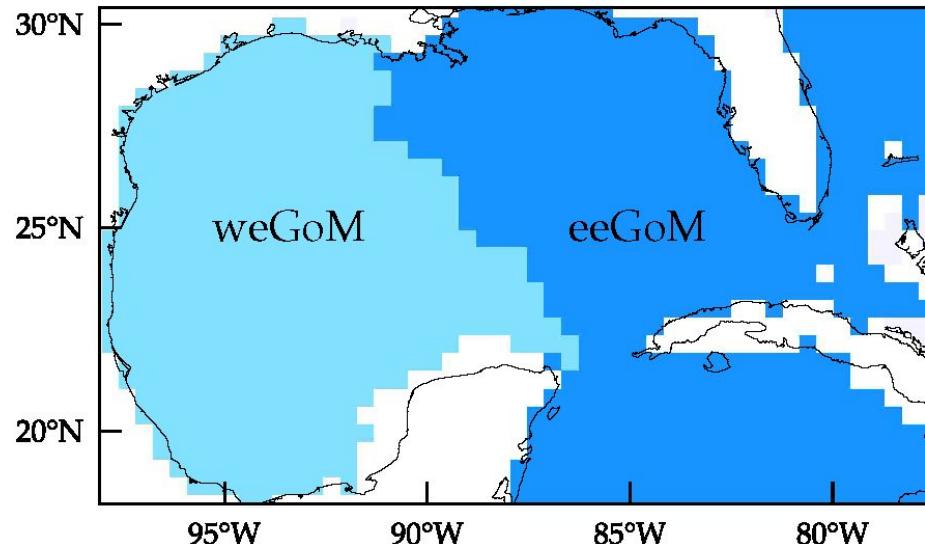
$$L_1^T = \begin{pmatrix} 0.83 \\ 0.55 \\ 0 \\ 0 \\ 0 \end{pmatrix} \quad L_2^T = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} \quad R_1 = \begin{pmatrix} 1 \\ 1 \\ 1 \\ 0 \\ 0 \end{pmatrix} \quad R_2 = \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \\ 1 \end{pmatrix}$$

- A, B are the attractors
- E is another attractor
- A, B, C basins of attraction
- D, E basins of attraction

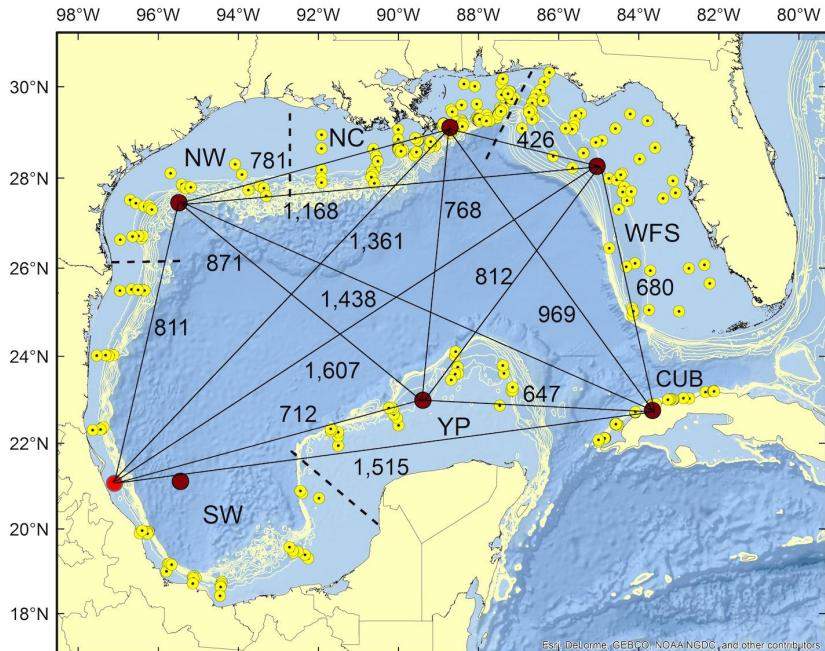
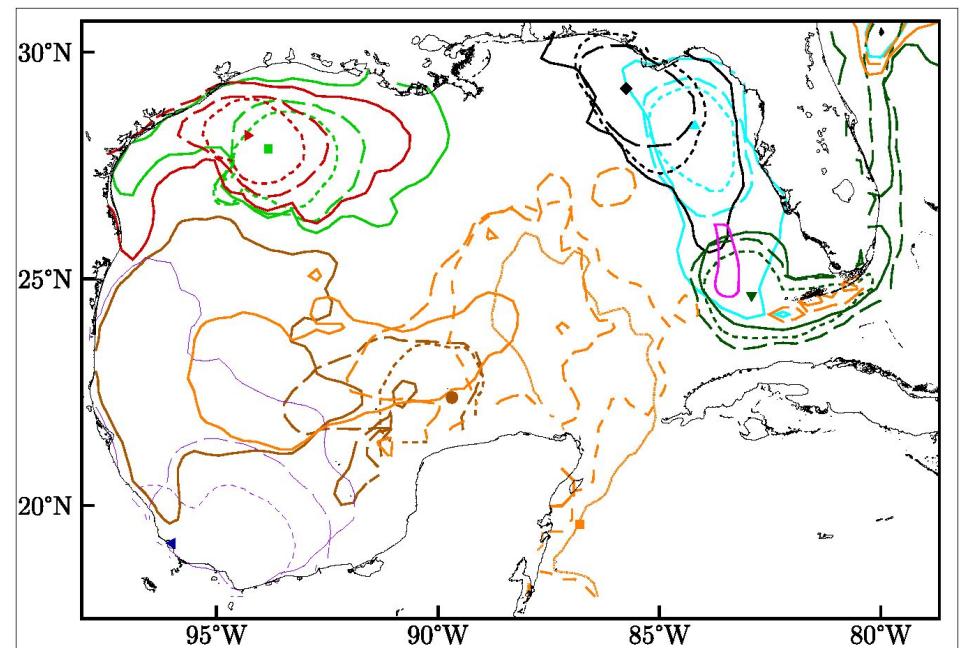
# Eigenvectors



# Lagrangian geography

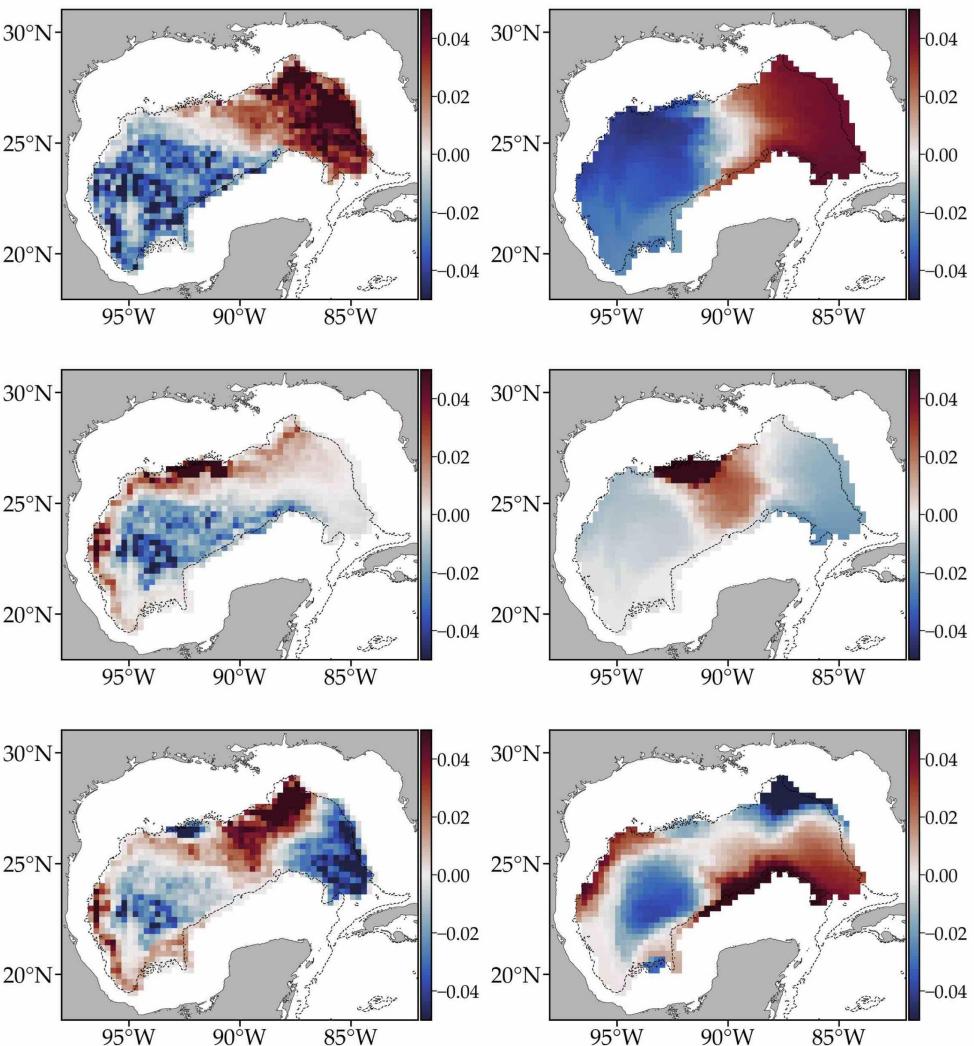


# Applications (Olascoa, 2018 and Murawski, USF)

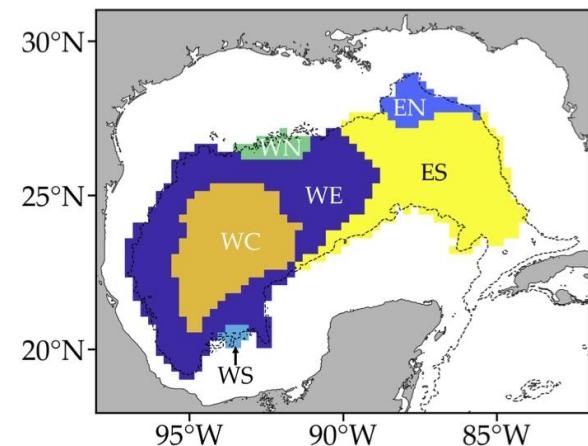
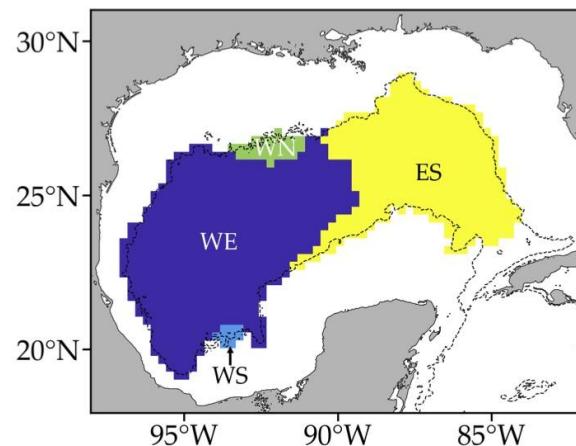
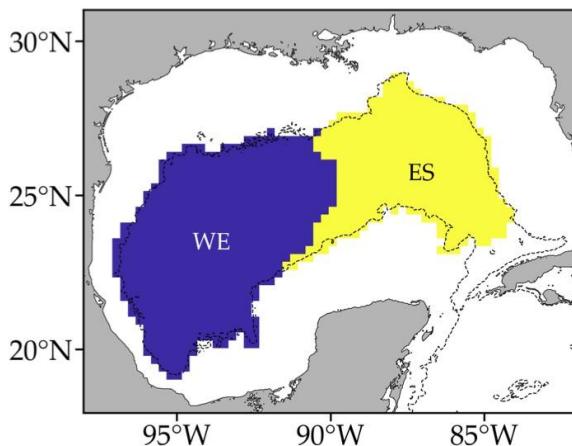


Connectivity of different Reefs in the GoM (left) and Group of tagged adult fish (right)

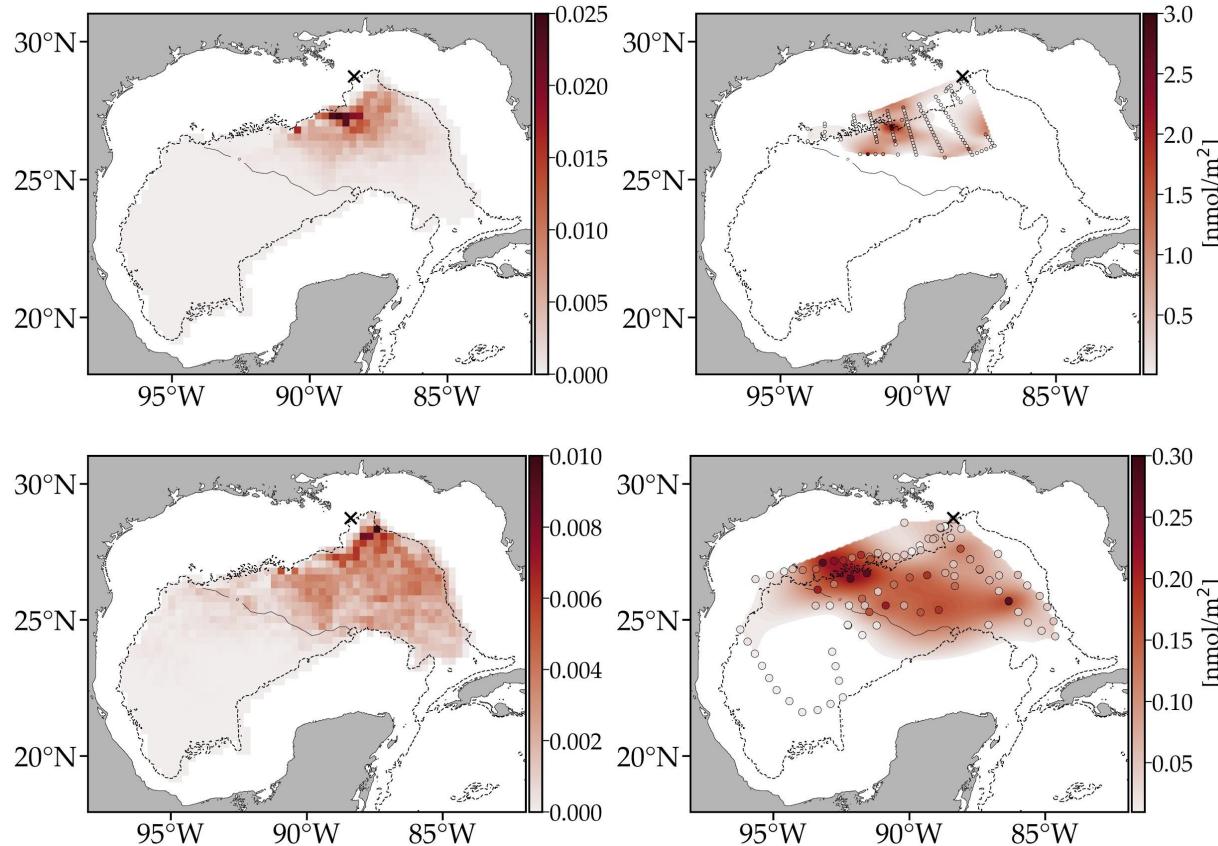
# Eigenvectors



# Lagrangian geography



# Comparison with tracer dispersion (Ledwell 2012)



# Markov-chain-inspired search for MH370

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LAPCOD, Venezia, 2019

# Catastrophe

The disappearance in the Indian ocean of Malaysian Airlines flight MH370 is one of the biggest aviation mysteries.

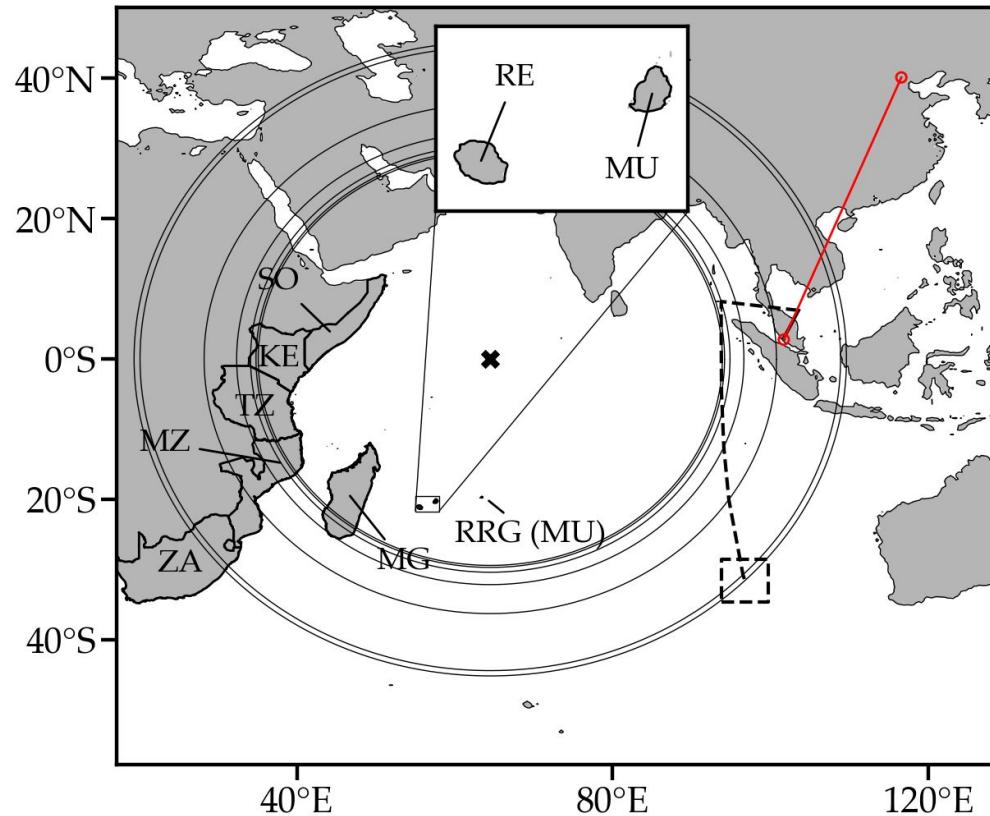
- second deadliest incident involving a Boeing 777 (227 passengers and 12 crew members)
- most expensive search in aviation history (\$155 million)
  - 2 years and 4 months
  - 4 vessels equipped with deep-water vehicles, sonars and cameras
  - underwater scan of a 120 000 km<sup>2</sup> search area

# Time of March 8, 2014 (UTC time)

- 16:42: takeoff from KUL to PEK with an expected arrival 22:30 (00:42–06:30 MYT)
- 17:06: aircraft final transmission (position and remaining fuel)
- 17:19: last verbal communication from the pilot to air traffic control
- 17:22–18:22: aircraft is observed by civilian and military radar
- until the transponder stopped working (or turned off)
- 19:41–00:19: seven log-on acknowledgements between the
- plane and the satellite
- 01:15: aircraft did not respond to a status request

23:24 (07:24 MYT): Malaysia Airlines stated that contact with the plane was lost at 17:21 and that search operations are launched.

# Path reconstruction from satellite *ping* signal

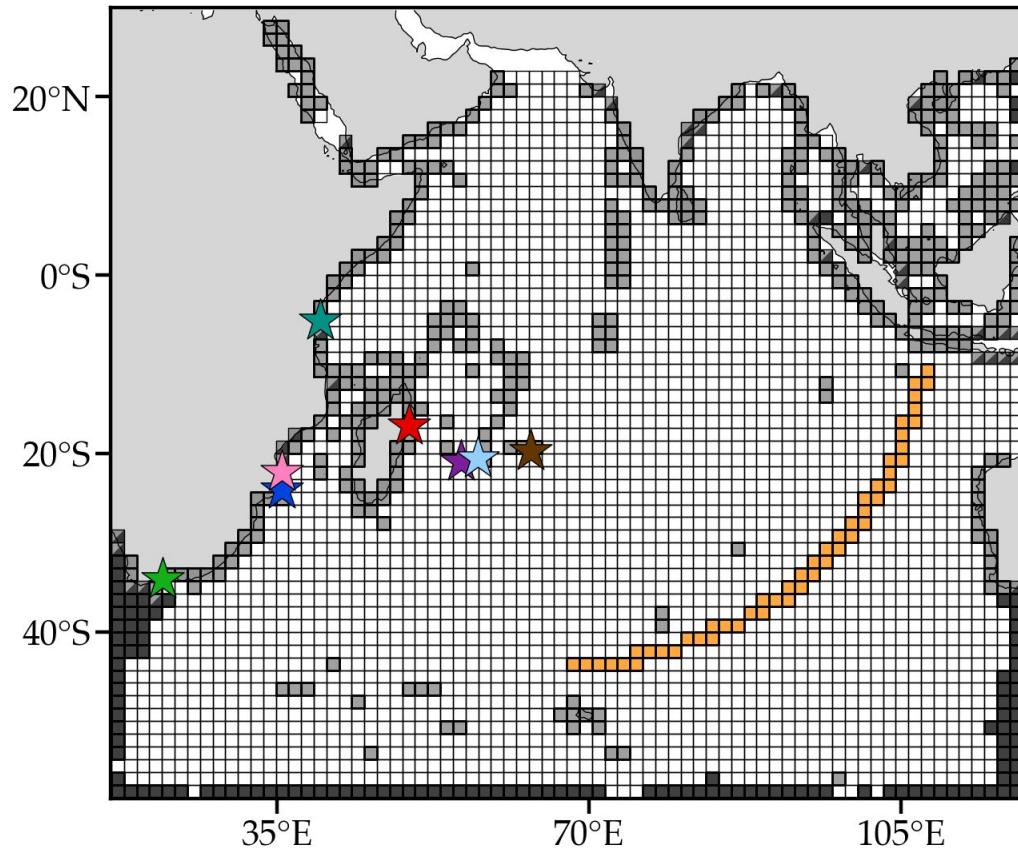


# First debris reached Reunion Island



508 days after the crash of the airplane

# Confirmed debris beaching (508–838 days)



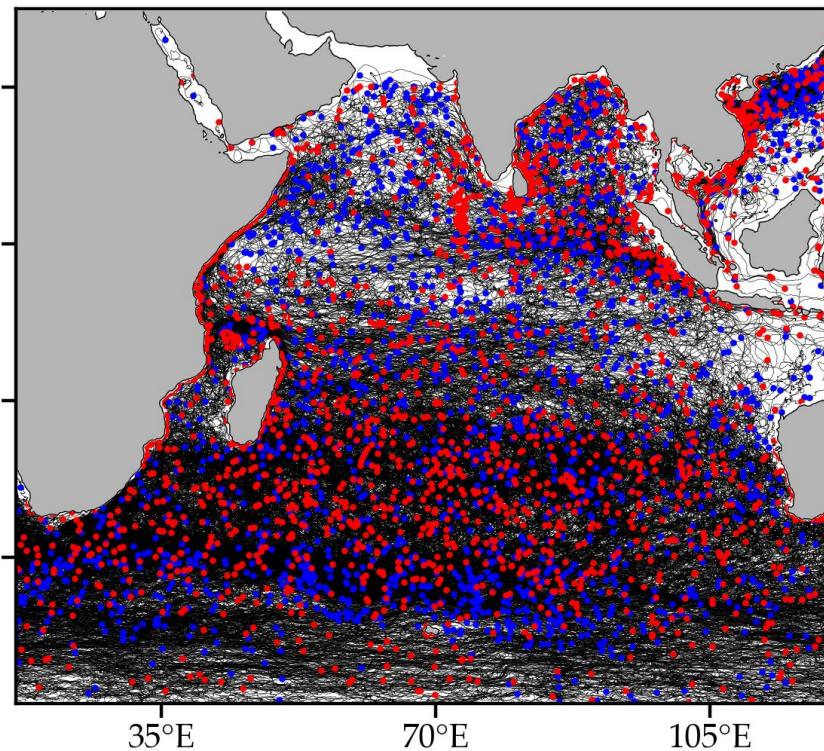
# Goal

Using undrogued drifters, which are subject to similar dynamics as floating debris, build a model to estimate the airplane crash site.

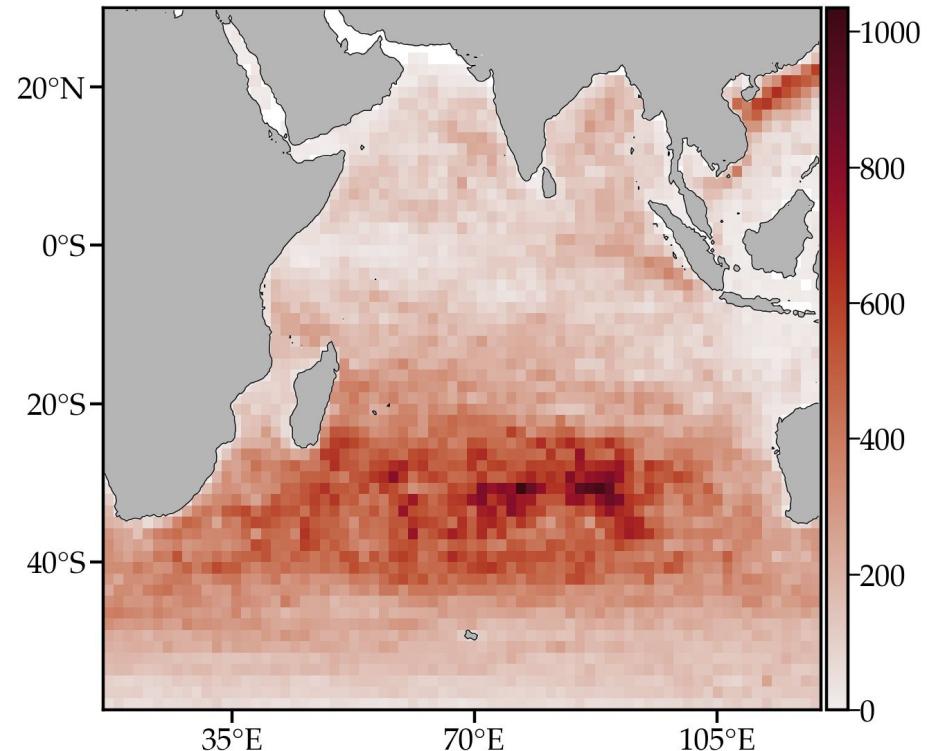
Model will utilize all available information:

- Last known satellite arc
- Debris beaching time and location

# Undrogued drifter in the Indian Ocean



Trajectoires since 1979 (left) and density of 226 data points per bin (right)



# Modelisation and adaptation

- Transition time:  $T = 5$  days
- Resolution :  $2.5^\circ \times 2.5^\circ$
- Seasonality of Monsoon : three transition matrices (summer, winter, spring/fall)
- Mass exiting the domain (to a nirvana states)
- Beaching probability equals to the land ratio

# Long-term behavior

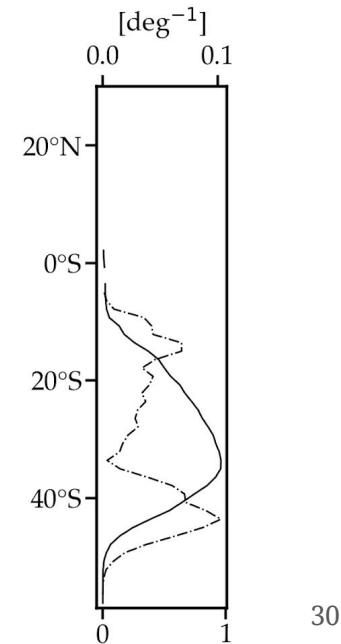
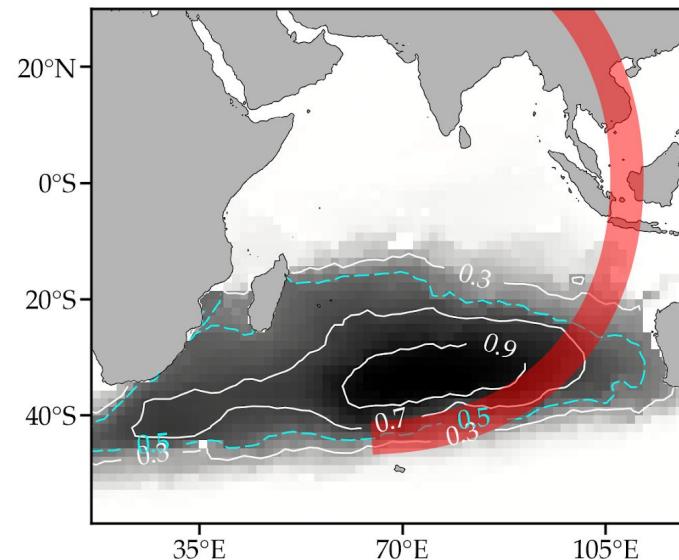
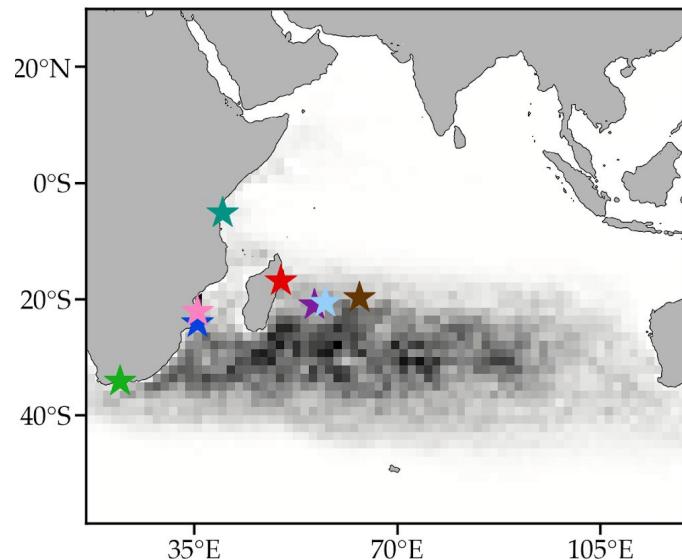
Long-term behavior can be observed from the spectral analysis of the autonomous season-aware matrix:

$$P = P_W^{18} \cdot P_{SF}^{18} \cdot P_S^{18} \cdot P_{SF}^{18}$$

Note that :  $18 \cdot T = 90$  days  $\sim 3$  months

# Quasistationary distribution

First approximation of the search area is obtain from the eigenvectors of the combined seasonal P along the satellite arc between the cyan dashed contour ( $\sim 3600$  km).



# Bayesian analysis

Estimates the posteriori probability distribution of the crash site that matches the available observations (debris location, time, satellite arc).

# Probability to beach at the time of a debris

The time  $\tau$  and the probabilities  $p$  taken until hitting target set  $b$  for the first time from the crash site  $c$  (any state on the arc).

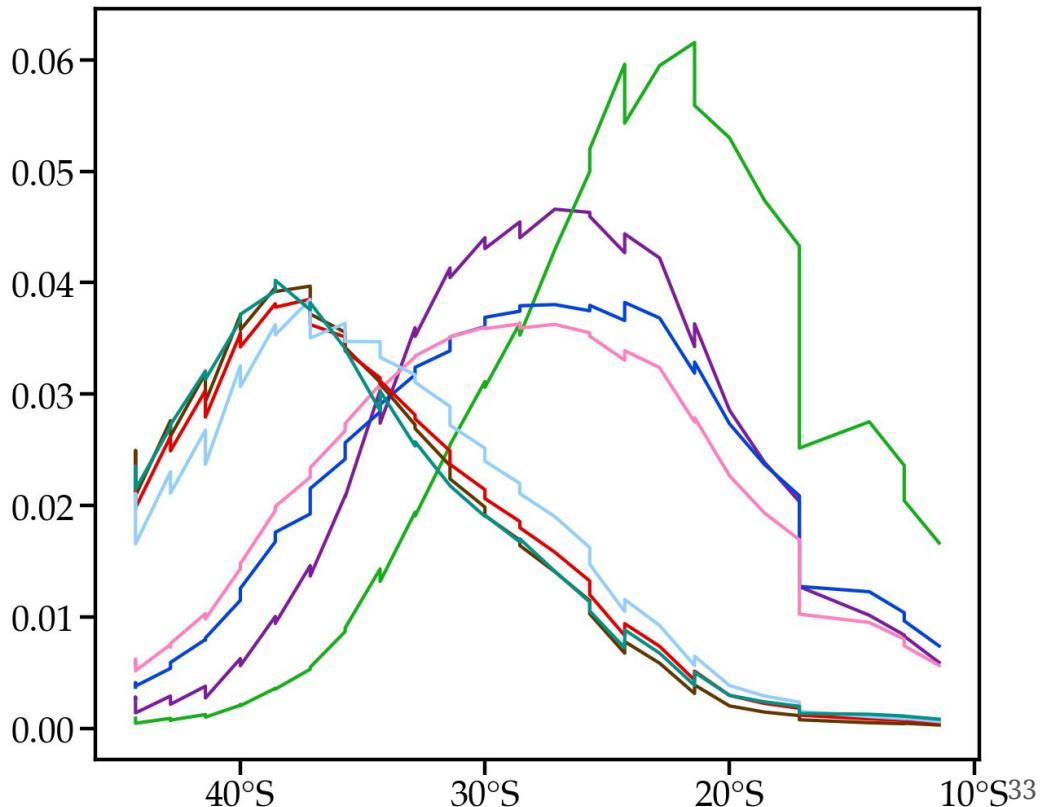
$$\tau^b = \inf_{k \geq 0} \{t + kT : \epsilon_{t+kT} = b\}$$

$$p_c^b(k) = \mathbf{1}_c P^k \cdot \mathbf{1}_b$$

$$\text{prob}[\tau^b = t+kT] = \begin{cases} p_c^b(k) = 0 & \text{if } k = 0 \\ p_c^b(k) - p_c^b(k-1) & \text{if } k > 0 \end{cases}$$

# Probability for each debris along the satellite arc

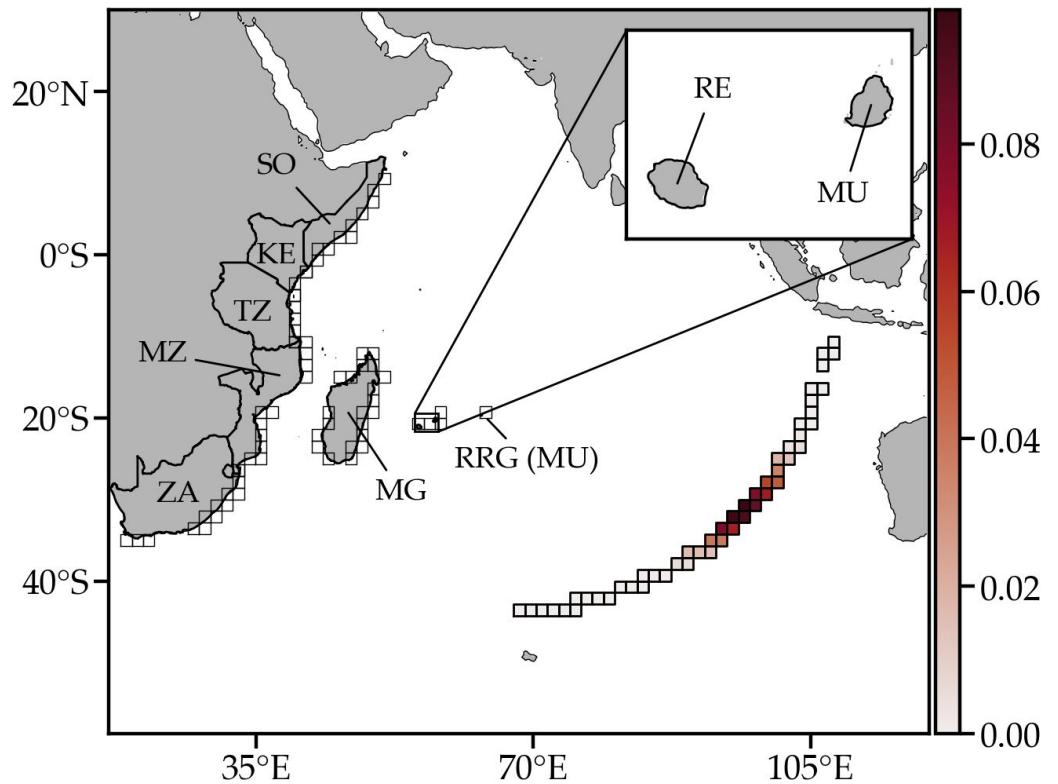
- Two groups ( $37^{\circ}\text{N}$  &  $25^{\circ}\text{N}$ )



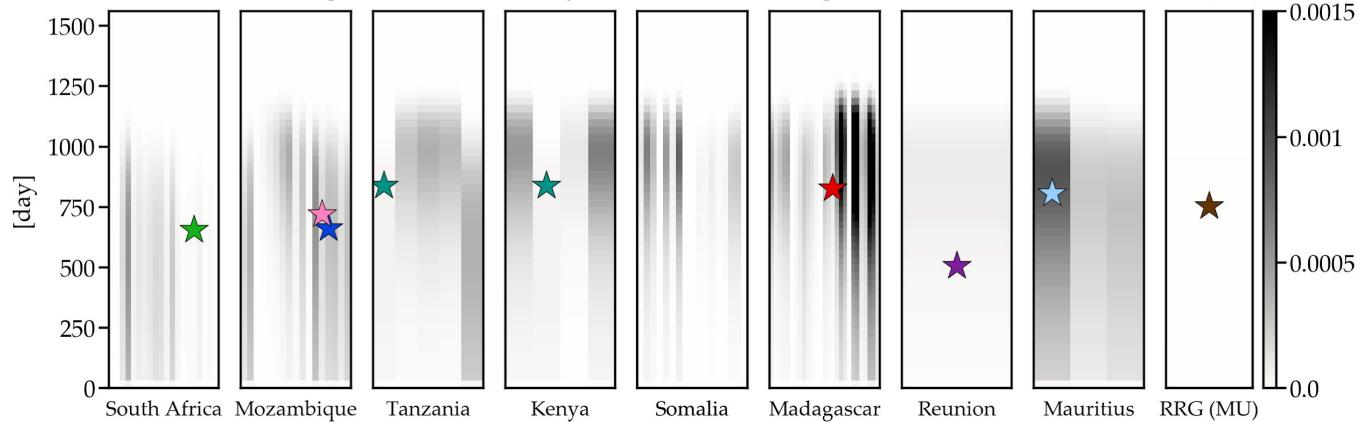
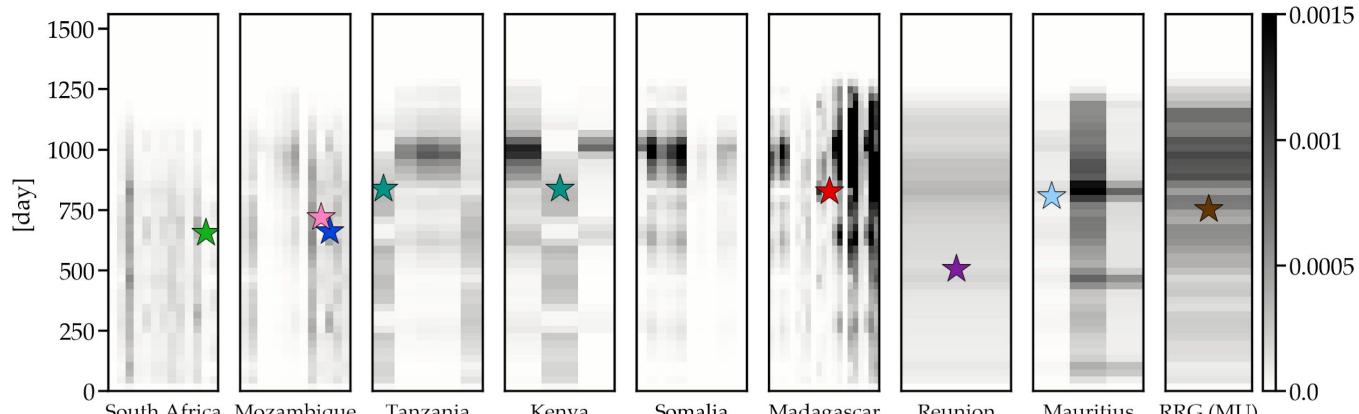
# Bayesian analysis

The posterior distribution is the product of all debris probability distributions.

$$p(t^b|c) = \prod_{b \in \mathcal{B}} p(t^b|c)$$



# Heatmap of the beaching probabilities

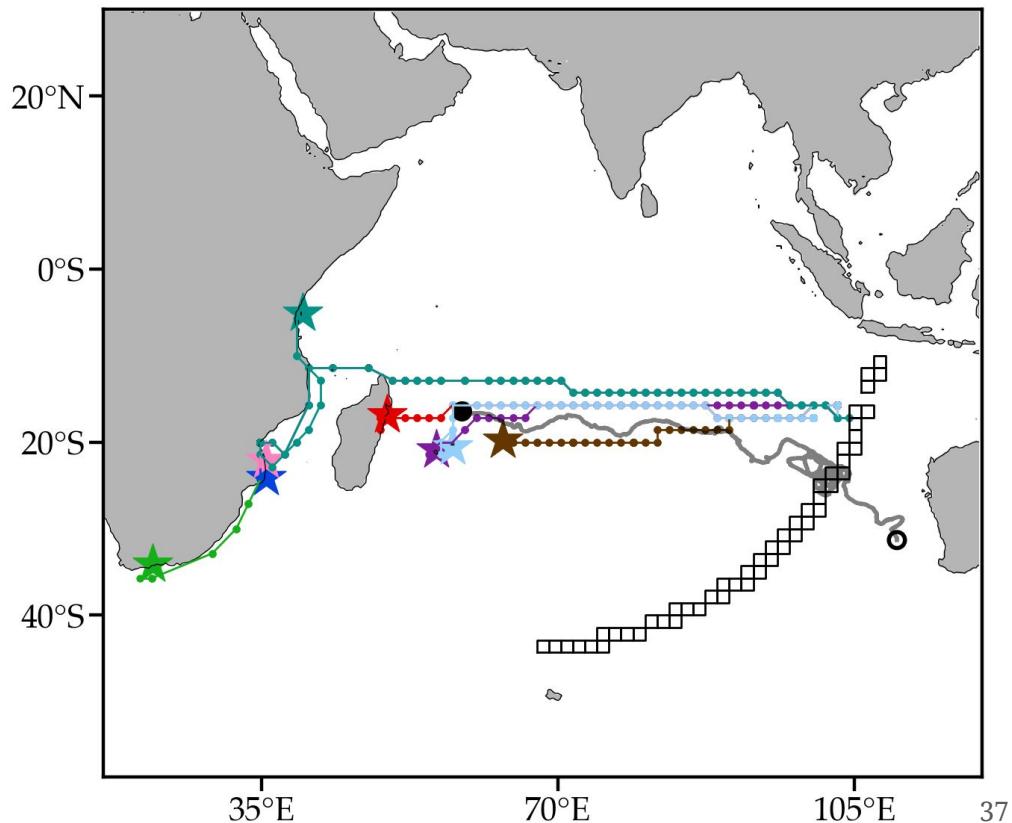


# Most probable path

1. Each debris (location and endtime)
2. Pushforward from probable initial location  $f(k=0)$  until endtime
3. Each step and each bin  $i$ , store:
  - a. The maximum probabilities ( $\max_j f(k-1)_j P_{ji}$ )
  - b. The bin  $j$  where this maximum probability comes from
4. On the last time step, same we have to reach the debris location.
5. Reconstruct the path backward

# Most probable paths

- Closer to north group
- Not too far from a GDP drifters present at the plane disappearance (Trianes, 2016)
- North of the search area
- Individual path have very low probability of reaching the target at the right time



# Future work

- *NSF project* : Nonlinear Dynamics Investigation of North Atlantic Deep Water Pathways (Beron-Vera, PI)
- Oil dispersion and evaporation (Olascoaga's & Sheinbaum's talk)