

# A Lagrangian geography of the deep Gulf of Mexico

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# Introduction

Over the last twenty five years, many satellite-tracked surface drifters sampled the surface of the Gulf of Mexico (GoM). From a database of 3300 drifters, we presented a Lagrangian geography of the GoM.

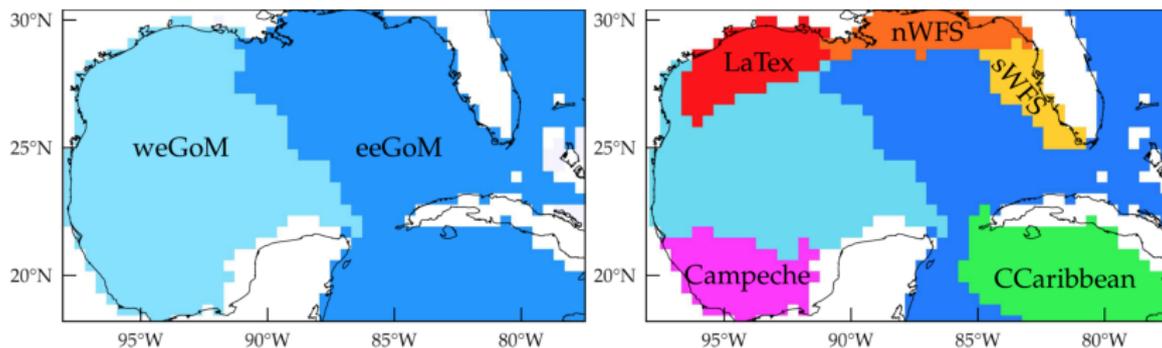


Figure: Published in [miron2017lagrangian](#)

# Introduction

In contrast, very few studies were aimed at the characterization of the deep water global circulation.

RAFOS experiments sponsored by the Bureau of Ocean Energy Management (BOEM) (July 2011 - May 2015) courtesy Alexis Lugo Fernandez:

- ▶ 121 floats at 1500 m
- ▶ 31 floats at 2500 m
- ▶ 4-year mission (floats  $\sim$  2-y mission and are redeployed)

# Objectives

Using floats data (trajectories) in the abyssal Gulf of Mexico (GoM):

- ▶ subdivide the deep GoM into regions with similar dynamics;
- ▶ identify sinks and sources (attractors and basins of attraction);
- ▶ predict transport of *passive tracers*.

## Theory: flowmap $F$

Given the drifter paths, we wish to access the map  $x \mapsto F(x)$  that determines how the drifters change positions.

If the ocean flow was stationary in some statistical sense, then the velocity could be expressed as  $v(x)$ . In this case, we could simply solve the ODE  $\dot{x} = v(x)$  to get  $F$ , the flow map, and *move around* any function, such as a tracer distribution  $f(x)$ .

## Theory: transfer operator $\mathcal{P}$

This is done by composing  $f$  with  $F^{-1}$  and it defines a transfer operator  $\mathcal{P}$  as the linear operator such that

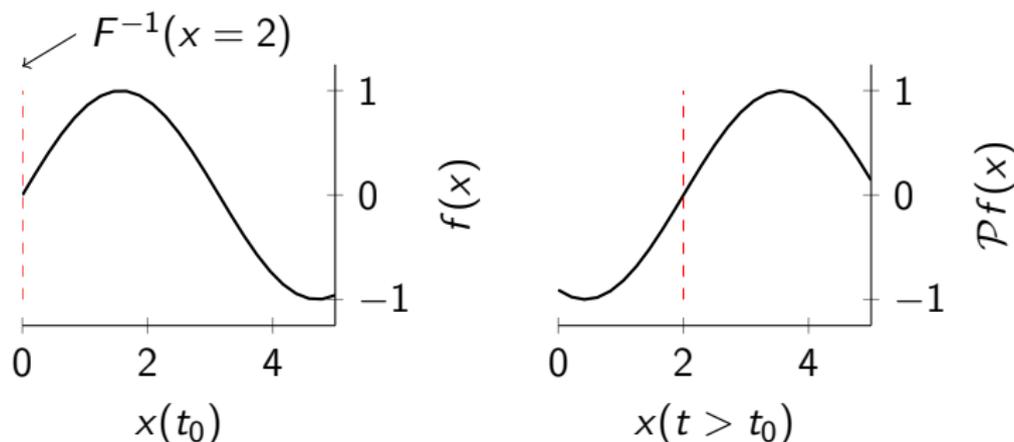
$$\mathcal{P}f(x) = f \circ F^{-1}(x). \quad (1)$$

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Why  $F^{-1}$ ? Example, if  $F$  moves the distribution 2 units to the right.



## Theory: transfer operator $\mathcal{P}$

For  $F$  representing a non-autonomous flow as above,  $\mathcal{P}f(x)$  is the solution of

$$\partial_t \rho + v \cdot \nabla \rho = 0, \quad \rho(x, t = 0) = f(x). \quad (2)$$

In our case,  $F$  is unknown and we only have access to trajectories. We will use a discrete method to approximate  $\mathcal{P}$ .

## Theory: transition matrix

Partition the domain into equal-size bins  $\{B_1, \dots, B_N\}$  (a regular grid) and consider the indicator function of set  $B_i$ :

$$\mathbf{1}_{B_i}(x) = \begin{cases} 1 & \text{if } x \in B_i, \\ 0 & \text{if } x \notin B_i. \end{cases} \quad (3)$$

Note that  $V_N := \{\mathbf{1}_{B_1}(x), \dots, \mathbf{1}_{B_N}(x)\}$  is a discrete orthonormal basis w.r.t. the inner product

$$\langle \bullet, \mathbf{1}_{B_i}(x) \rangle = \frac{1}{\text{area}(B_i)} \int_{B_i} \bullet \mathbf{1}_{B_i}(x) d^2x = \overline{\bullet}^{B_i}. \quad (4)$$

Then, project  $f(x)$  on the basis  $V_N$ , called Ulam basis (**ulam1960**):

$$\Pi_N f(x) = \sum_1^N f_i \mathbf{1}_{B_i}(x), \quad f_i = \langle f(x), \mathbf{1}_{B_i}(x) \rangle = \overline{f(x)}^{B_i} \quad (5)$$

where  $\Pi_N$  is the projector.

## Theory: how to construct the transition matrix

In a similar manner one can project  $\mathcal{P}$  on  $V_N$ :

$$\begin{aligned}(\Pi_N \mathcal{P})_{ij} &= \frac{1}{\text{area}(B_j)} \int_{B_j} \mathcal{P} \mathbf{1}_{B_i}(x) \cdot \mathbf{1}_{B_j}(x) d^2x \\ &= \frac{\text{area}(B_i \cap F^{-1}(B_j))}{\text{area}(B_i)} =: P_{ij}.\end{aligned}\tag{6}$$

The entries of  $P$  can be viewed as transitional probabilities of moving from  $B_i$  to  $B_j$  (Markov Chain with bins  $\equiv$  states):

$$P_{ij} = \frac{\# \text{ of particles in } B_i \text{ that are mapped to } B_j}{\# \text{ of particles in } B_i}.\tag{7}$$

## Theory: transition matrix

$P_{ij}$  gives us the action of  $F$  at a coarse-grained level given by the partition. This introduces diffusion proportional to the size of  $B_i$  and is solution of:

$$\partial_t \rho + v \cdot \nabla \rho = D \nabla^2 \rho \quad \text{with } \rho(x, t = 0) = f(x), \quad (8)$$

with  $D \propto \text{area}(B_i)$ .

## Application of the transition matrix

One can push forward discrete representations of  $f(x)$ :

$$\mathbf{f} = (f_1, \dots, f_N), \quad (9)$$

by right-multiplication by  $P$ :

$$\begin{aligned} f^{(1)} &= f P \\ f^{(2)} &= f^{(1)} P = f P^2 \\ f^{(k)} &= f P^k \end{aligned} \quad (10)$$

Similarly, we push backward an initial distribution using  $P^\top$ .

## Eigenvectors analysis

It is also of interest to identify when a distribution  $\mathbf{f}$  is almost invariant:

$$\mathbf{f} \approx \mathbf{f} P \quad (11)$$

This is available from the *eigenspectrum* inspection of  $P$  (Froyland et al., 2012).

If in the matrix  $P$ :

- ▶ all states *communicate*;
- ▶ no state occurs *periodically*.

$P$  has a limiting distribution  $\mathbf{p} = \mathbf{p}P$ .

Note:  $\mathbf{p}$  is a left eigenvector of  $P$  with eigenvalue  $\lambda = 1$  ( $\mathbf{p}\lambda = \mathbf{p}P$ ). Because of row-stochasticity of  $P$ , the corresponding right eigenvector is  $\mathbb{1}$ , i.e.,  $P\mathbb{1} = \mathbb{1}$ .

## Attractors and basin of attractions

Any distribution  $f_{\mathbb{1}}$  supported on the right eigenvector  $\mathbb{1}$  will converge to  $\mathbf{p}$  as the number of applications of  $P$  tends to infinity, i.e.,  $\lim_{n \rightarrow \infty} f_{\mathbb{1}} P^n = \mathbf{p}$ .

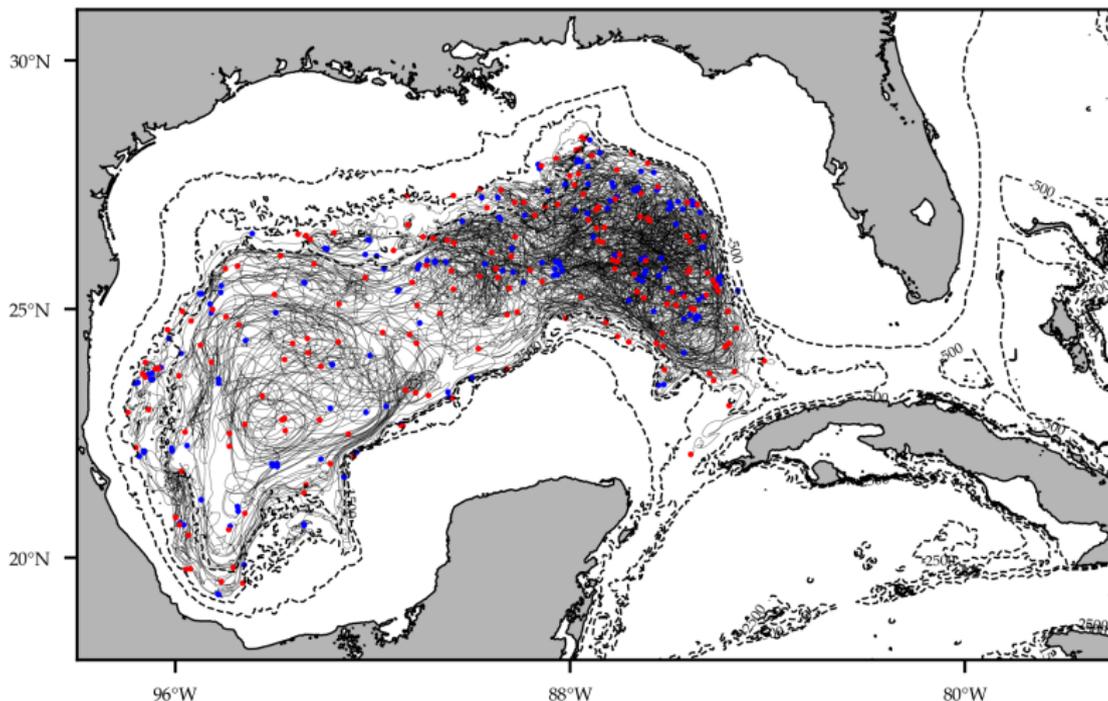
For  $\lambda = 1$ :

- ▶ right eigenvector of  $P$  is the basin of attraction
- ▶ left eigenvector of  $P$  is the attractor

This motivates the idea that regions where trajectories converge and their basins of attraction are encoded in the eigenvectors of the transition matrix  $P$  with eigenvalues ( $\lambda \approx 1$ ) (**froyland2014well**).

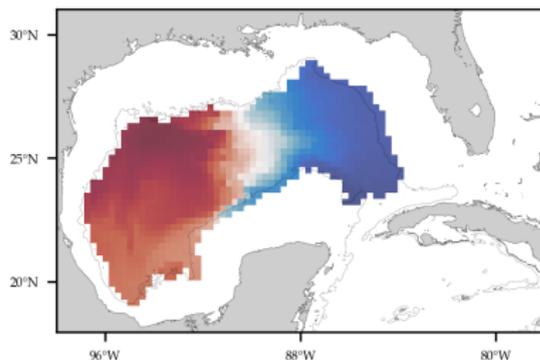
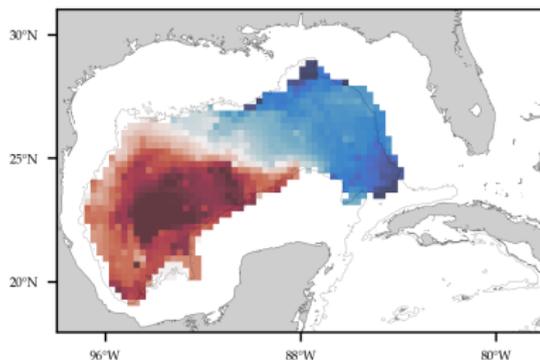
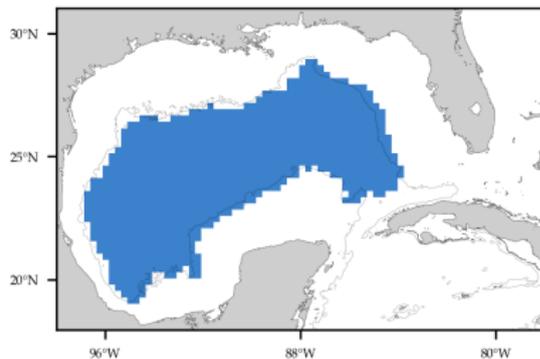
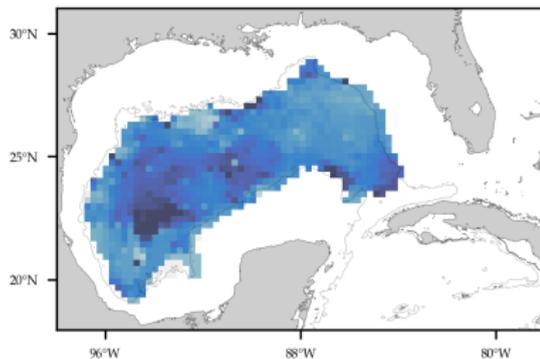
# Data

Trajectories of the RAFOS cover the area under 1500 m of the Gulf of Mexico.



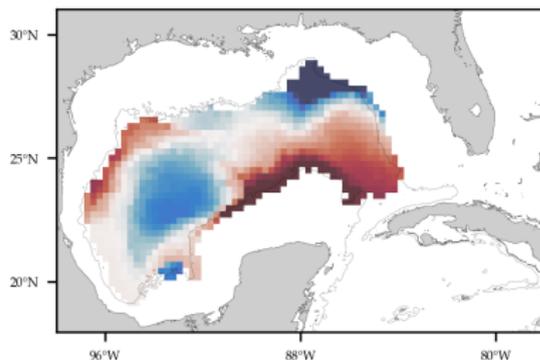
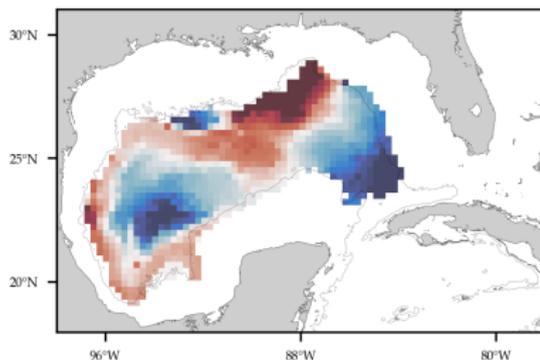
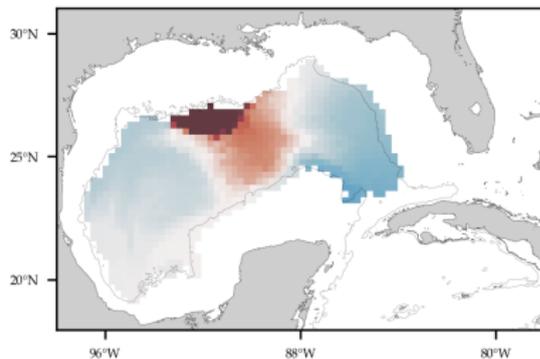
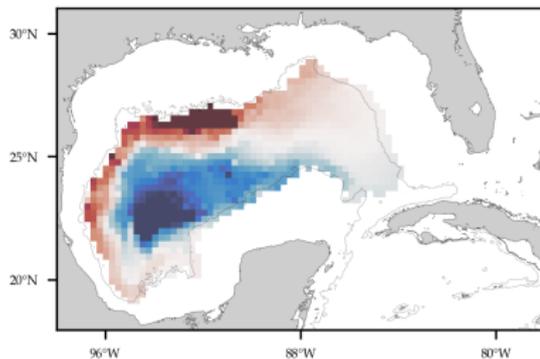
# Eigenvectors

Eigenvectors associated with  $\lambda_1 = 1$  and  $\lambda_2 = 0.9953$ .



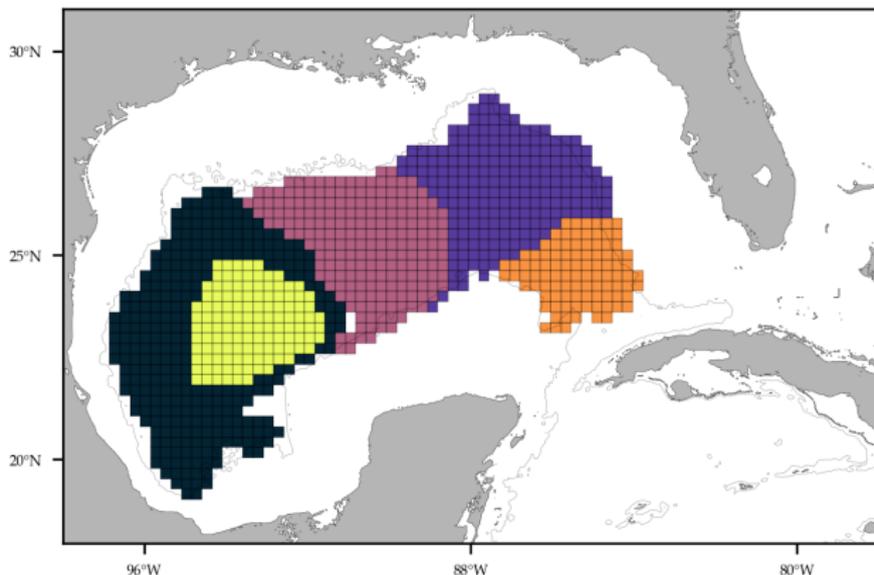
# Eigenvectors

Eigenvectors associated with  $\lambda_3 = 0.9832$  and  $\lambda_5 = 0.9712$ .



## Lagrangian geography of the deep Gulf of Mexico

Combination of the basins of attraction from the top right eigenvectors (by thresholding).

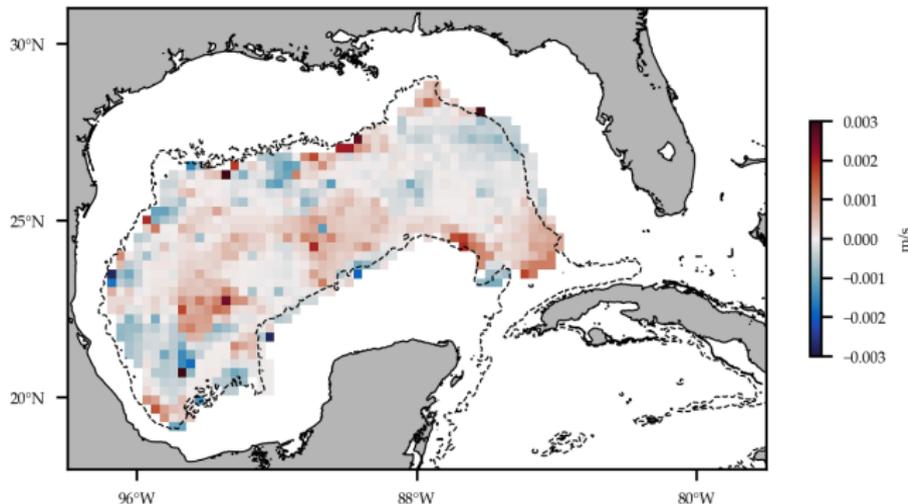


Residence timescale: 7-y left (black) and right (purple), 1-y middle (pink), 0.5-y gyre (yellow) and bottom right (orange)

## Upwelling and Downwelling

From incompressibility, accumulation can be used to approximate vertical velocity  $w$ :

$$h_{t_1} = h_{t_0} \frac{\text{area}_{t_0}}{\text{area}_{t_1}} \quad w \sim \frac{h_{t_1} - h_{t_0}}{dt}. \quad (12)$$



The vertical component is about  $\sim 1\%$  of the maximum horizontal velocity ( $u_{max} = 0.2739\text{m/s}$ ,  $v_{max} = 0.1790\text{m/s}$ ).

# Comparison with experimental data (ledwell2016dispersion)

Tracer evolution from the infamous Deepwater Horizon oil spill.

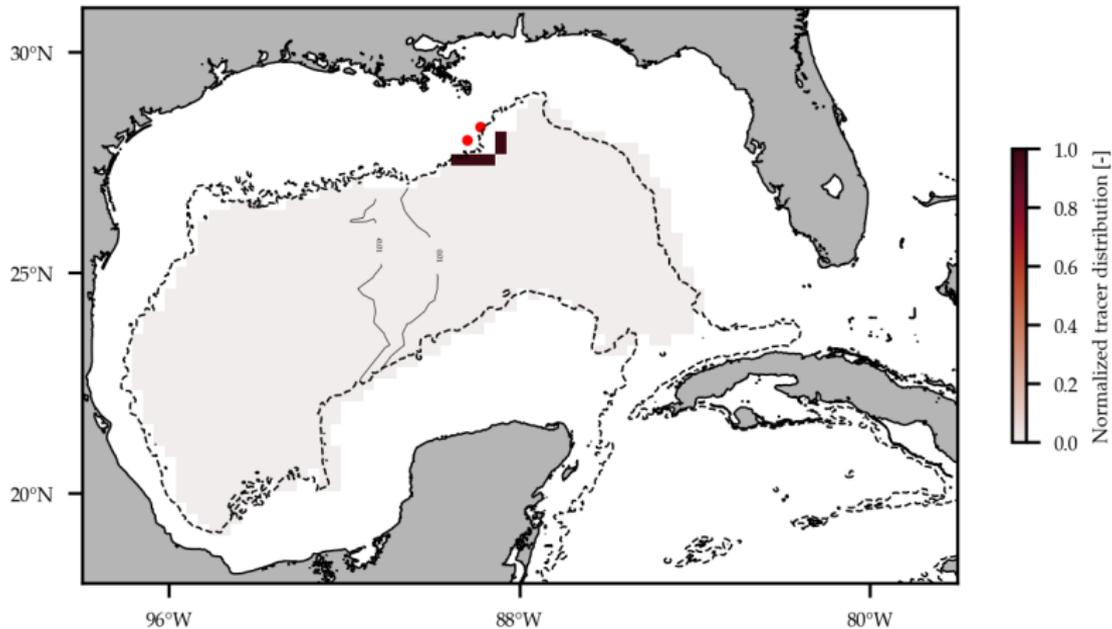


Figure: Initial location of the tracers

# Comparison with experimental data (ledwell2016dispersion)

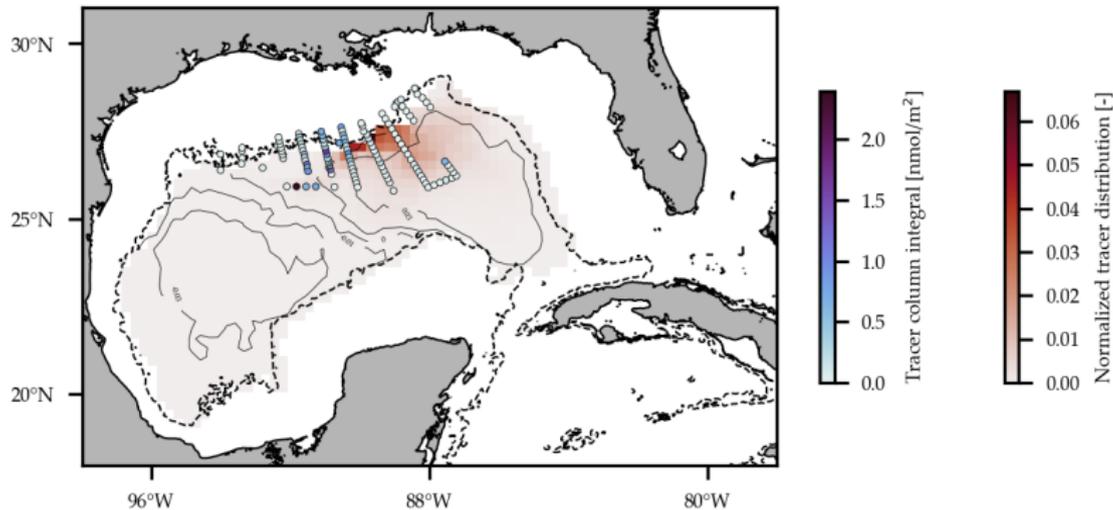


Figure: Evolution after 4 months

# Comparison with experimental data (ledwell2016dispersion)

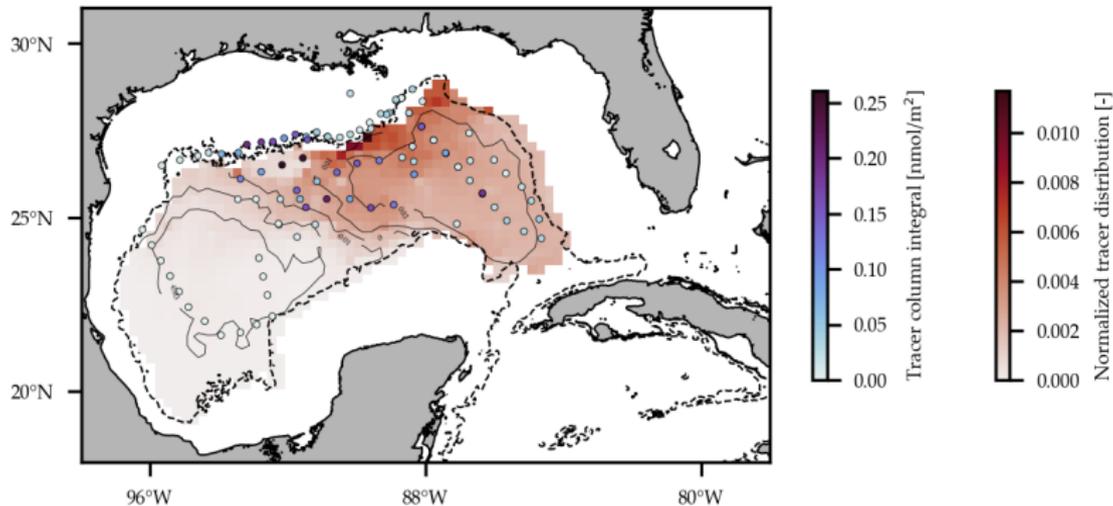


Figure: Evolution after 12 months

# Thank you!

Future plans:

- ▶ plastic sources and convergence zones
- ▶ take account inertial effects (see next talk!)

Any questions?

## References I

Froyland, G. et al. (2012). “Three-dimensional characterization and tracking of an Agulhas Ring” . In: *Ocean Modelling* 52-53, pp. 69–75, 2012.