A dynamical geography of the Gulf of Mexico

Philippe Miron¹, Francisco J. Beron-Vera¹, María J. Olascoaga², Paula Pérez-Brunius³, Julio Sheinbaum³ and Gary Froyland⁴

¹RSMAS/ATM University of Miami, Miami, USA

²RSMAS/OCE University of Miami, Miami, USA

³CICESE, Ensenada, Mexico

⁴University of New South Wales, Sydney, Australia

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Using only drifters data (trajectories) in the Gulf of Mexico (GoM):

- Subdivide the GoM into regions with similar dynamics
- Identify sink and source (attractor and basins of attraction)
- Predict transport of passive and possibly non-passive tracers

Drifters database in the GoM (1994-2016)

Problematics: non uniform data (data per year, type of drifters, time resolution)

- Left: initial positions
- Right: all trajectories data points



3312 drifters from different sources (LASER & GLAD / CARTHE, GDP / NOAA, BOEM / SCULP, PEMEX / CICESE)

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Biodegradable drifters (Novelli et al., 2017, University of Miami)

- 1000 drifters deployed during the LAgrangian Submesoscale ExpeRiment (LASER) in 2016
- Total height of 0.6 m
- ▶ GPS precision ±10 m
- Quarter-hourly acquisition (~ 3 months)



Approximation of an attractor (Dellnitz and Junge, 1997)



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Transfer operator

- ► Domain X
- Map $T: X \circlearrowleft$ acts on a density $f: X \to \mathbb{R}$



Transfer operator

For all the points $\in X$, if the map T is area preserving, the end result can be obtained from the Perron-Frobenius operator (P):

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



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Transition matrix (Froyland, 2001)

- Probabilistic approach known as Ulam's method (Ulam, 1960)
- Subdivision of the domain in boxes (B_1, \cdots, B_n)
- From short-time trajectories, we approximate the probability to go from box i to box j:

$$\mathcal{P}_{ij} pprox rac{\#\{p: p \in B_i ext{ and } T(p) \in B_j\}}{\#\{p \in B_i\}}$$

by counting the number of particles (drifters) in B_i that are mapped into B_j .

Markov Chain and eigenvectors method (Froyland, Stuart, and van Sebille, 2014)

- *P* defines a Markov Chain of the dynamics
- Using initial density $f_0 \rightarrow$ future distribution

$$f_1 = f_0 \mathcal{P}$$
$$f_N = f_0 \mathcal{P}^{\Lambda}$$

- left eigenvectors ($\lambda L = L \mathcal{P}$)
 - for $\lambda = 1$: invariant distribution
 - for $\lambda \approx 1$: almost invariant distribution
- right eigenvector highlights the basin of attraction

Example with a simple 5 states problem



Example with a simple 5 states problem



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

- L_1 : A, B are attractors
- ► *L*₂: *E* is another attractor
- R_1 : A, B, C basin of attr.

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• R_2 : D, E basin of attr.

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Hypothesis

All drifters trajectory start at the same time (Autonomous system) and we construct the transition matrix by looking where drifters end up 2 days later (bins size, data).



Algorithm

- 1. Split the domain (GoM) into square boxes
- 2. For each trajectory segment:
 - ▶ find bins *i* where x₀ is located and store the segment *ID* in the vector B_i
 - ▶ identify bins j where x_f is located and store the segment ID in the vector B_j

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- 3. Calculate the transition matrix \mathcal{P}_{ij} using vectors B_i and B_j
- 4. Calculate eigenvalues and eigenvectors of $\ensuremath{\mathcal{P}}$

Strongly connected components (Tarjan algorithm) The GoM is almost completely covered by a single CC.

- Borders: Closed Communicating Classes (CCC) in red with Attractive Closed Communicating Classes (ACCC) in black
- Handful of drifters and small effect on the general dynamics



Limiting distribution from a uniform density

Existence of a westward mean flow: similar to results presented by Sturges, 2016 from SSH data



Top left eigenvectors



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Top right eigenvectors



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Dynamical geography

- left: the main separation of the GoM
- right: the five coastal basins of attraction



Conclusion

- Identified almost limiting distributions and corresponding basins of attraction from the inspection of the eigenvectors
- Supported by independent observations of a westward mean flow (Sturges, 2016)
- Ability to push-forward a density to "predict" the dispersion (e.g. after an oil spill or to plan a drifters experiment)

Thank you!

Open questions:

- What is the influence of the drifters' type on the transition matrix ?
- Will it be possible to perform a seasonal (or maybe monthly) evaluation of the transition matrix ?
- How does it compare to high number of artificial drifters or simply density advection using a numerical velocity field or a global circulation model?

How to "easily" extract the different sets ?

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References II



