

Eddy identification from along track altimeter data using deep learning: EDDY project

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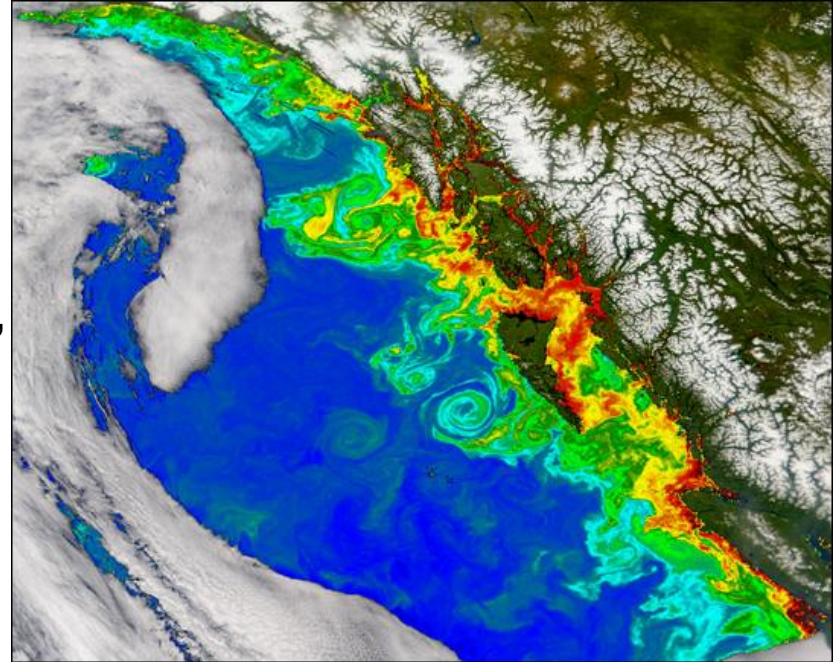


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Introduction

- Ocean eddies are cyclonically or anticyclonically rotating water masses.
- Eddies enable vertical transfer of heat, and also global scale heat transfer.
- Mesoscale eddies are of radius >100 km, with a lifespan weeks to months.
- Eddies appear to be highs or lows on Sea surface temperature (SST) and salinity (SSS), as well as sea surface height (SSH) anomaly maps
- Monitoring eddies is important for marine biologists and commercial fishery.



Source: <https://www.deepseanews.com/2012/09/an-ocean-full-of-eddies/>

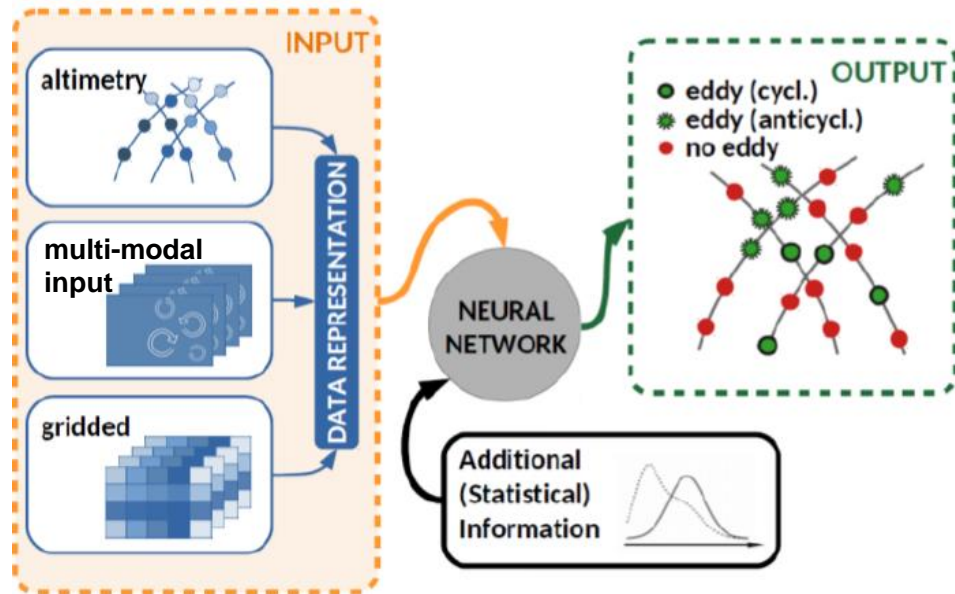
Eddy monitoring

- Eddy identification and tracking via SSH and SST
- For now, ONLY from gridded multi-mission SSH maps ([Chelton et al. 2007,2011](#); [Mason et al. 2014](#))!
- **Gridding inevitably lead to the loss of spatial (e.g. amplitude) and temporal resolution**
- Eddy identification can only be a delayed-time product with a lag of one to two months

EDDY project

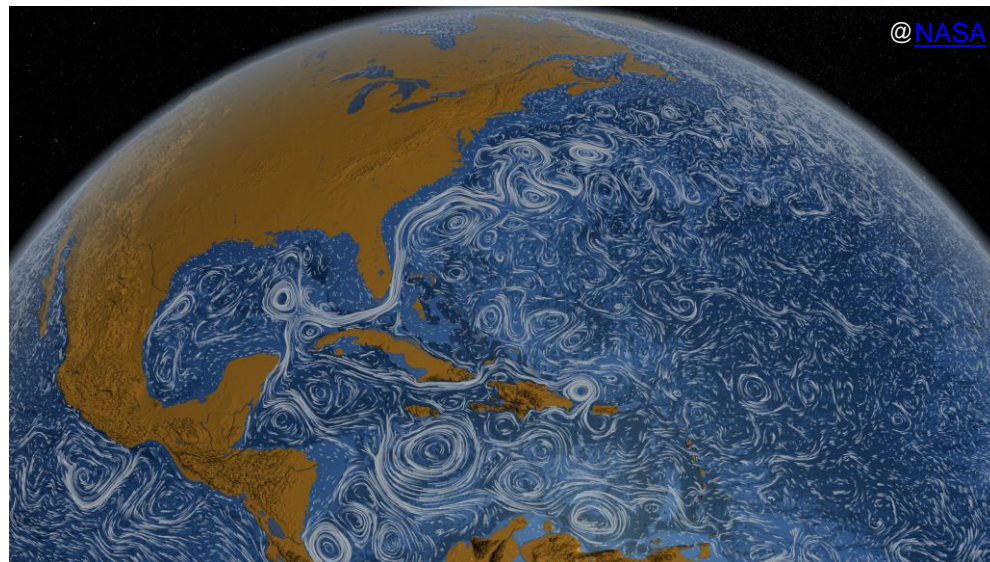


- Aim to develop eddy identification methods **directly from along track altimetry data**
- **Develop deep learning architecture** which integrate multi-modal remote sensing data like SSH and SST
- Near-real-time eddy classification and tracking may be achievable.
- Kick off in April 2021, with 3 year funding from [DFG](#).



Preliminary results

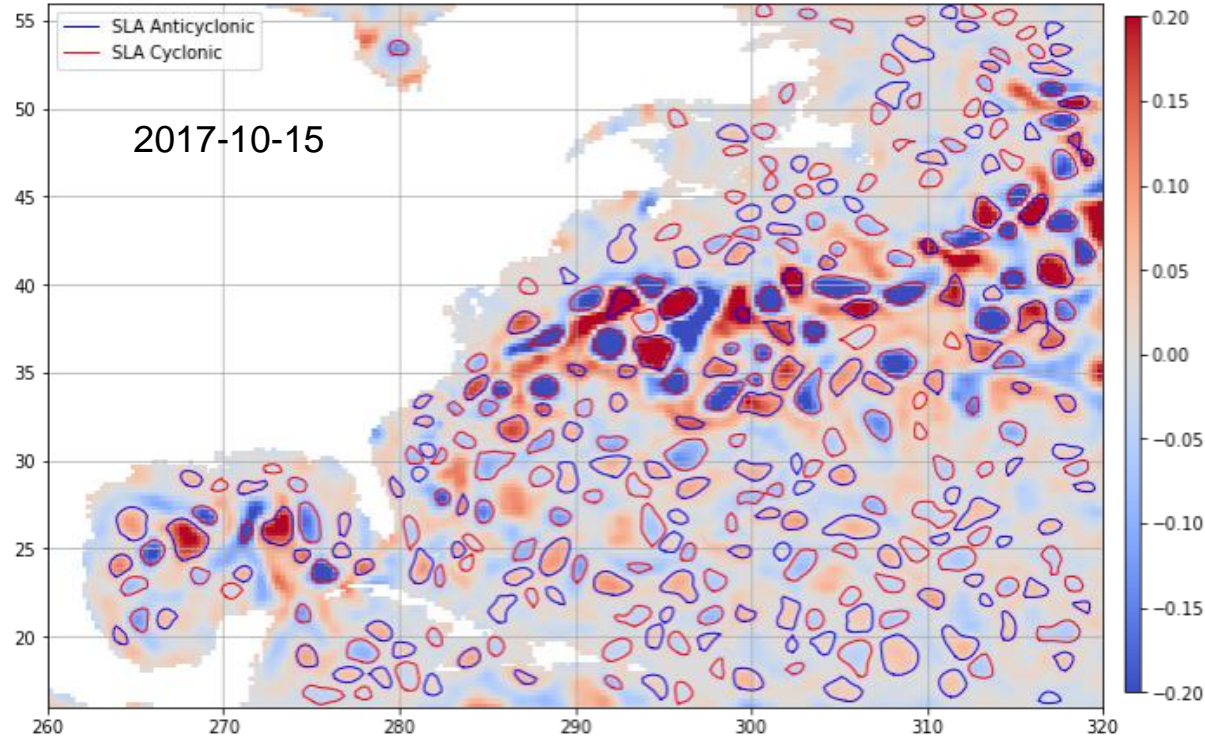
- Started with single-modal training with SSH, gridded maps
- py-eddy-tracker¹ ([Mason et al. 2014](#))
→ open source and fast eddy detection
- Study area in the Gulf Stream with one year of data (365 days)



Reference dataset

- Gridded datasets from CMEMS
- Eddy detection from py-eddy-tracker with and without Okubo-Weiss method

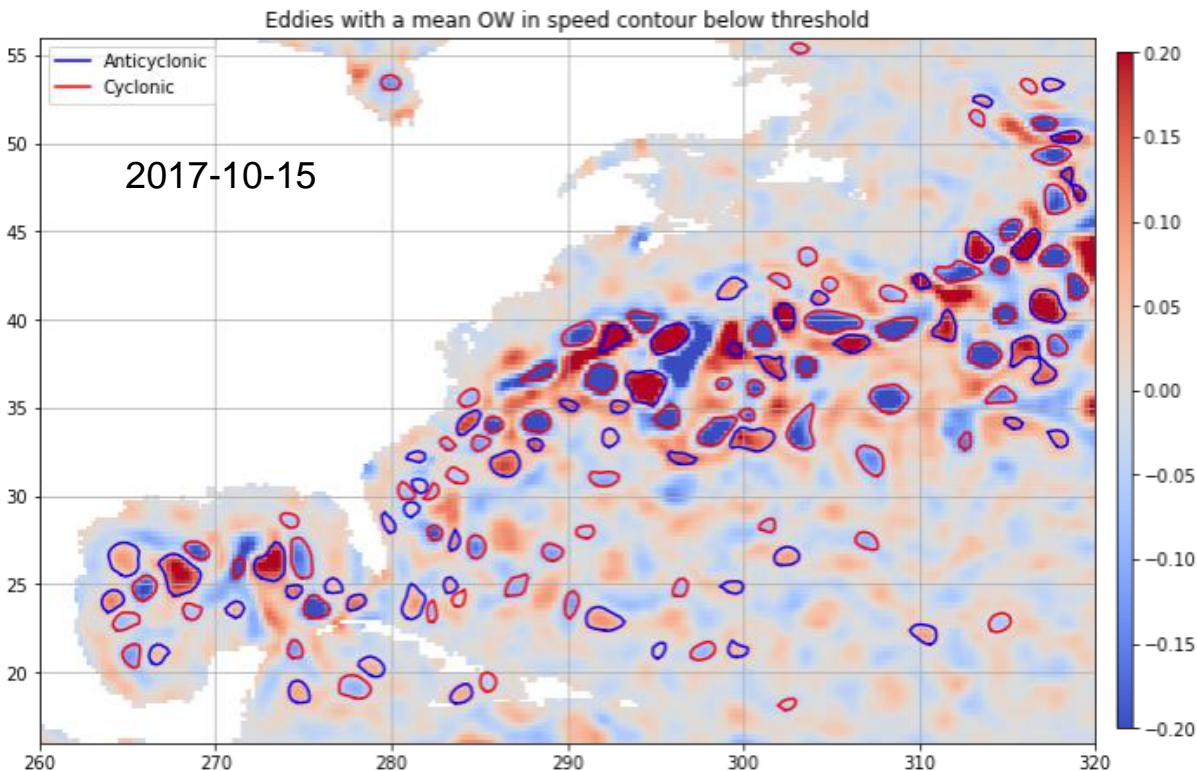
Eddy type	# of eddies
Anticyclonic (all)	142
Anticyclonic (with OW)	56
Cyclonic(all)	157
Cyclonic(with OW)	72



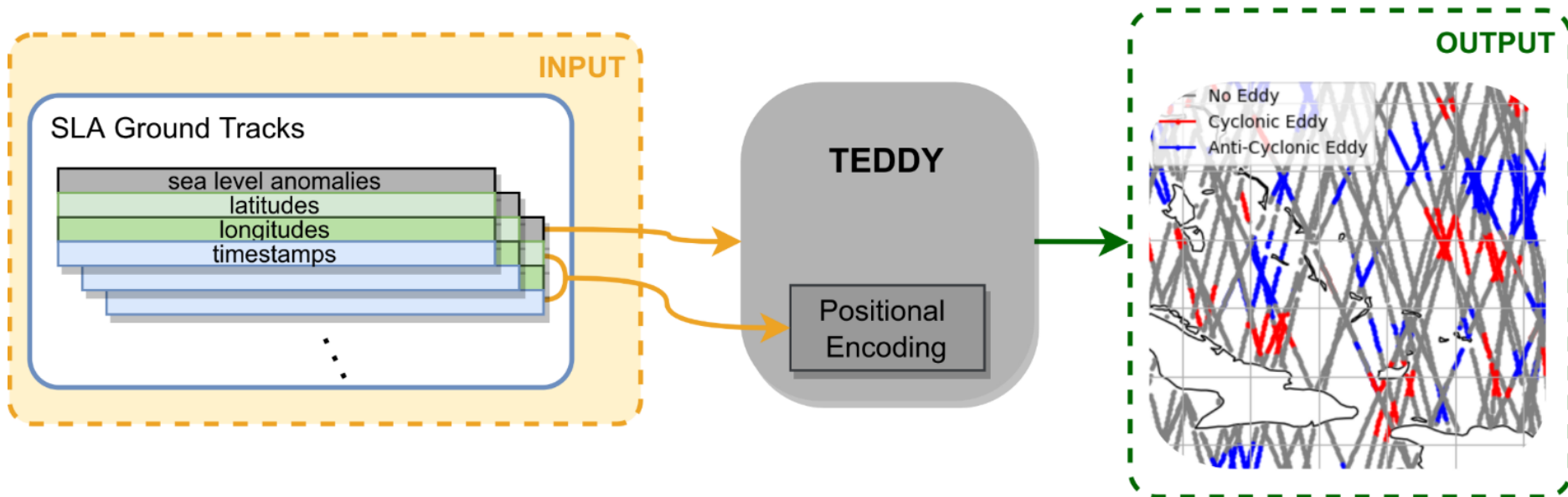
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TEDDY - Deep learning architecture



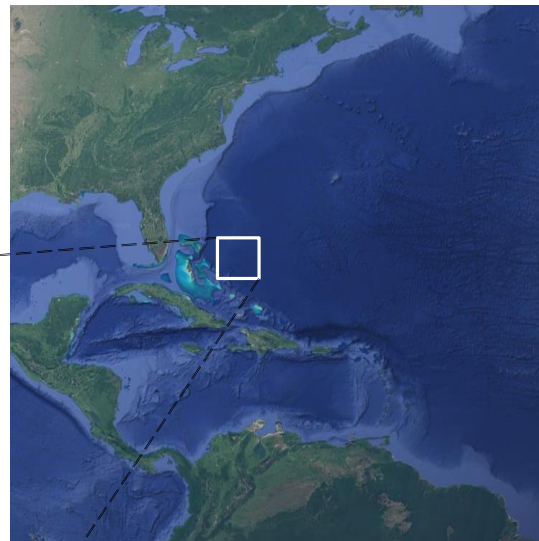
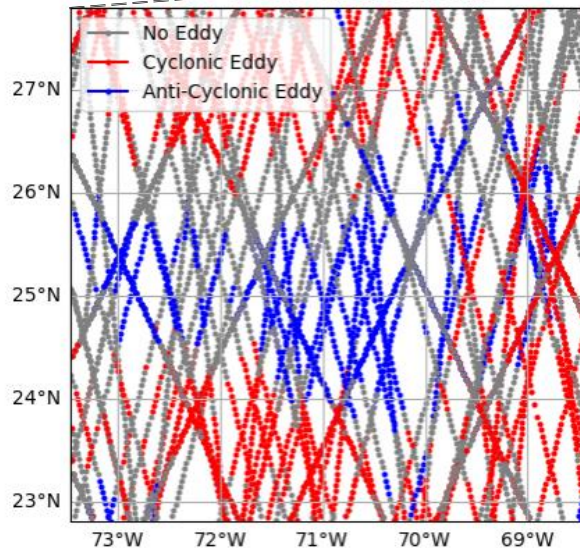
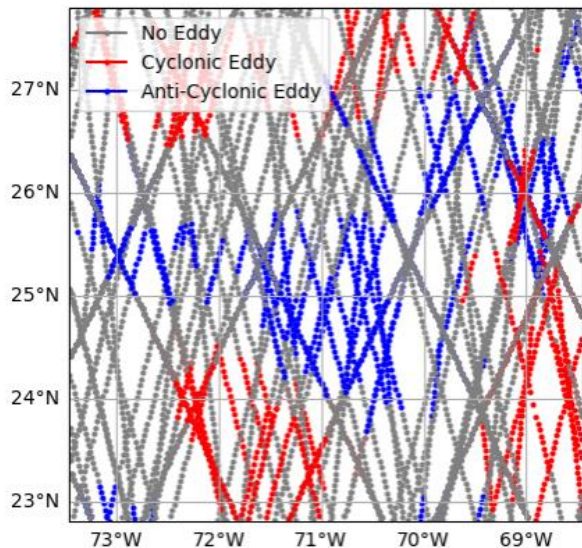
For details,
Re-visit

Eike et al., Session ITS2.7/AS5.2, Monday, 23 May, 16:15 - 16:20
[Machine learning-based identification and classification of ocean eddies](#)

TEDDY predictions

Reference

Prediction



Eddy identification from the first 40 days of 2017 multi-mission data

Remarks and outlook

- EDDY project aims to classify eddies directly from along track altimetry data.
- We aim to eventually classify and track eddies from ground tracks with a higher resolution and accuracy.
- We obtained promising results from single-modal training and eddy identification in the Gulf Stream.
- We adopt multi-modal training and classification in the next phase.
- SWOT mission data will be a valuable addition to the training set in 2023.

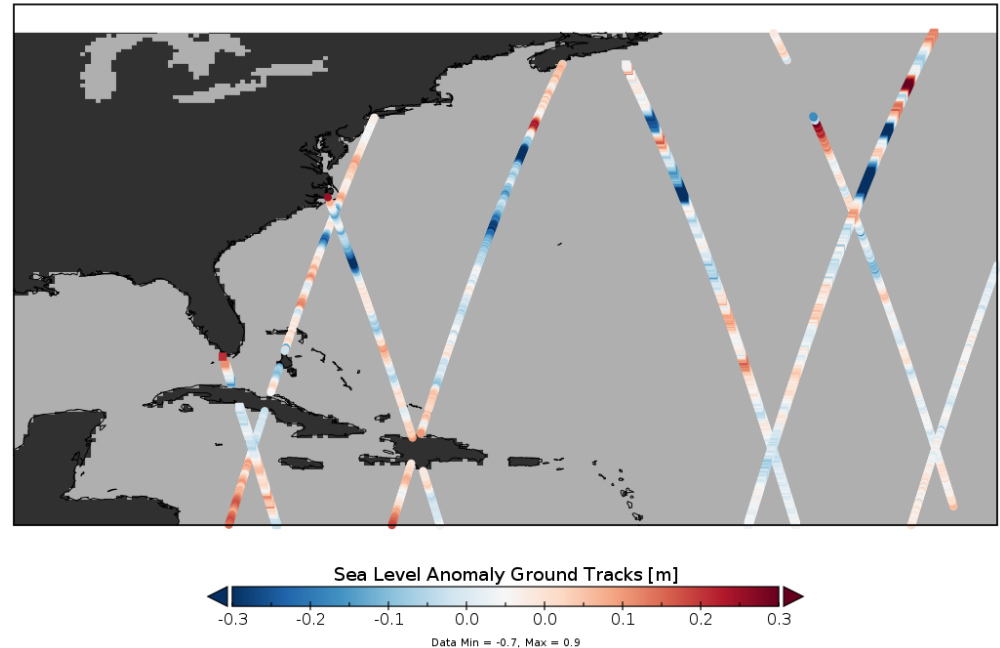
References

- Mason, E., Pascual, A., & McWilliams, J. C. (2014). A new sea surface height–based code for oceanic mesoscale eddy tracking. *Journal of Atmospheric and Oceanic Technology*, 31(5), 1181-1188.
- Chelton et al. (2007). Global observations of large oceanic eddies. *Geophys. Res. Lett.*, 34, L15606.
- Chelton et al. (2011). Global observations of nonlinear mesoscale eddies. *Prog. Oceanogr.*, 91(2), 167-216.
- Okubo (1970). Horizontal dispersion of floatable particles in the vicinity of velocity singularities such as convergences. *Deep Sea Res. Oceanogr. Abstr.*, 17(3), 445-454.
- Weiss (1991). The dynamics of enstrophy transfer in two-dimensional hydrodynamics. *Physica D*, 48(2), 273-294.
- Franz, K., Roscher, R., Milioto, A., Wenzel, S., and Kusche, J. (2018). Ocean eddy identification and tracking using neural networks. In *IEEE IGARSS 2018* (pp. 6887-6890).
- Drees, L., Kusche, J., and Roscher, R. (2020). Multi-Modal Deep Learning with Sentinel-3 Observations for the Detection of Oceanic Internal Waves, *ISPRS Ann. Photogramm. Remote Sens. Spatial Inf. Sci.*, V-2-2020, 813–820.
- Lguensat et al. (2018). EddyNet: A deep neural network for pixel-wise classification of oceanic eddies. In *IEEE IGARSS 2018* (pp. 1764-1767).

Classification from along track altimetry - Input Data

- Sea Level Anomaly Ground Tracks (Level 3) from multiple missions from CMEMS
- 01.01.2017 - 31.12.2017
- Processed reference data using the py-eddy-tracker product of 2D grid data

Ground Track Sample from Saral/AltiKa Mission provided by CMEMC



EDDY project concept

A deep learning neural network learns and integrates along track and gridded altimetry dataset along with other multi-modal inputs like SST and multi-spectral images

