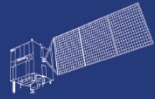


HY



HJ-1AB



CBERS



Gaofen



Beijing-2



Sentinel-1



Sentinel-2



Sentinel-3



Sentinel-5p



Aeolus

2023 DRAGON 5 SYMPOSIUM

3rd YEAR RESULTS REPORTING

11-15 SEPTEMBER 2023

PROJECT ID. 57192

REMOTE SENSING OF CHANGING COASTAL
MARINE ENVIRONMENTS (RESCCOME)

WEDNESDAY, 12 SEPTEMBER 2023, S.2.1: COASTAL ZONES AND OCEANS

ID. 57192

**PROJECT TITLE: REMOTE SENSING OF CHANGING COASTAL MARINE ENVIRONMENTS
(RESCCOME)**

PRINCIPAL INVESTIGATORS: MARTIN GADE & LI XIAO-MING

**CO-AUTHORS: MERETE BADGER, CHARLOTTE HASAGER, ABDALMENEM OWDA,
SEBASTIAN PETERS, SIMON SCHÄFERS, WANG WENSHENG, ZHANG DI,
ZI NANNAN**

PRESENTED BY: MARTIN GADE

Objectives:

- Exploitation of Copernicus Sentinels, ESA, ESA TPM and Chinese EO data: better understanding of the ways, in which **coastal ecosystems** are exposed to, and react on, **various anthropogenic impacts**
- Scientific exchange: bi- or multi-lateral research or educational activities, **joint Sino-European research packages**
- Publication of co-authored results: **joint publications** at the Midterm and Final Dragon 5 Symposia, and in leading peer-reviewed scientific journals
- Training to young European and Chinese scientists: webinars, **excursions**, educational courses, and **summer schools**

Contribution of the Partners:

Research Packages:	European partners					Chinese partners				
	UHH	UoA	UoB	UiT	DTU	AIRCAS	OUC	NSOAS	HNTOU	TJU
57192-1: Intertidal regions	Teamleader				Teamplayer	Teamplayer				Teamleader
57192-2: Offshore wind farms					Teamleader	Teamleader				
57192-3: Offshore oil pollution	Teamplayer	Teamplayer		Teamleader			Teamleader	Teamplayer	Teamplayer	
57192-4: Coastal pollution		Teamleader	Teamplayer	Teamplayer				Teamleader	Teamplayer	
57192-5: Coastline changes	Teamplayer	Teamplayer	Teamleader			Teamplayer	Teamplayer		Teamleader	Teamplayer

Cross-Cutting Themes:

Synergism of RS data	Teamplayer	Teamplayer	Teamplayer	Teamleader	Teamplayer	Teamplayer	Teamleader	Teamplayer	Teamplayer	Teamplayer
Processing of Big Data		Teamleader		Teamleader	Teamplayer	Teamleader	Teamleader		Teamplayer	
Coastal stress factors	Teamleader	Teamleader				Teamleader			Teamleader	Teamplayer
Education of Young Scientists	Teamplayer	Teamplayer	Teamleader	Teamplayer	Teamplayer	Teamplayer	Teamplayer	Teamplayer	Teamleader	Teamplayer
Dissemination and outreach	Teamleader	Teamplayer	Teamleader	Teamplayer	Teamplayer	Teamplayer	Teamplayer	Teamleader	Teamplayer	Teamplayer

Teamleader	Teamleader
Teamplayer	Teamplayer
involved	involved

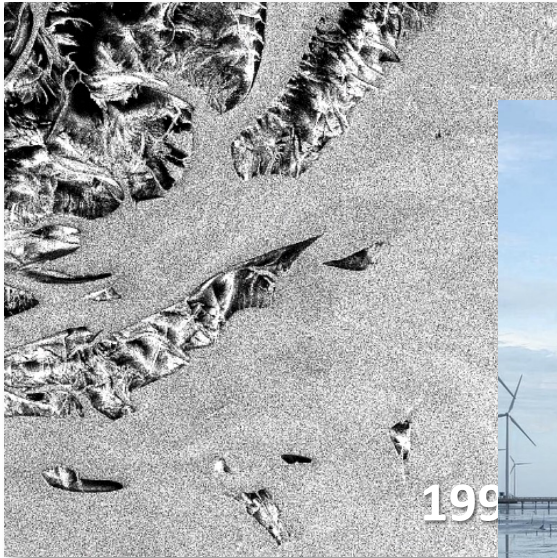
ESA /Copernicus Missions	No. Scenes	ESA Third Party Missions	No. Scenes	Chinese EO data	No. Scenes
1. Sentinel-1	>1000	1. TerraSAR-X	>10	1. GF-1	400 PMS 400 WFV Cam
2.		2. ALOS	5	2. Gf-3	12 Strip Dual
3.		3. Radarsat-2	2	3. GF-6	400 WFV Multi
4.		4.		4. SDGSAT-1	>500 GLI >2600 TIS
5.		5.		Total:	>4300
6.		6.		Issues:	
Total:		Total:	>17	SDGSAT-1 (launched Nov 2021) freely available: https://www.sdgsat.ac.cn/	
Issues:		Issues:		GF-1, GF-3, and GF-6 data available: https://data.cresda.cn/	
Through ESA's Sentinel Hub and from local archives at UHH and DTU.		From local archives at UHH.			

Name	Institution	Poster title	Contribution including period of research
Simon Schäfers	UHH	A Neural Network for the Detection of Water Lines	MSc student, since 2022
Sebastian Peters	UHH	N/A	BSc student, since 2021
Abdalmenem Owda	DTU	N/A	PhD student, since 2021

Name	Institution	Poster title	Contribution including period of research
Zhang Di	UHH	Classification of Intertidal Flat Surfaces by Means of Deep Learning	PhD student, 2020-2022
Zi Nannan	AIRCAS	Oceanic Eddy Detection from SAR Imagery Based on Deep Learning Network	PhD student, since 2021
Qiu Yujia	AIRCAS	Retrieval of Sea Ice Drift in the Arctic Based on Sequential Sentinel-1 SAR Data	PhD student, since 2020
Huang Bingqing	AIRCAS	N/A	PhD student, 2019-2023
Jia Tong	AIRCAS	N/A	PhD student, 2018-2022

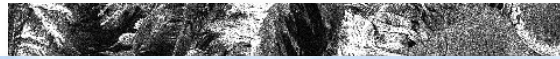
ReSCCoME: Remote Sensing of Changing Coastal Marine Environments

Longterm changes – example: Chinese coast:

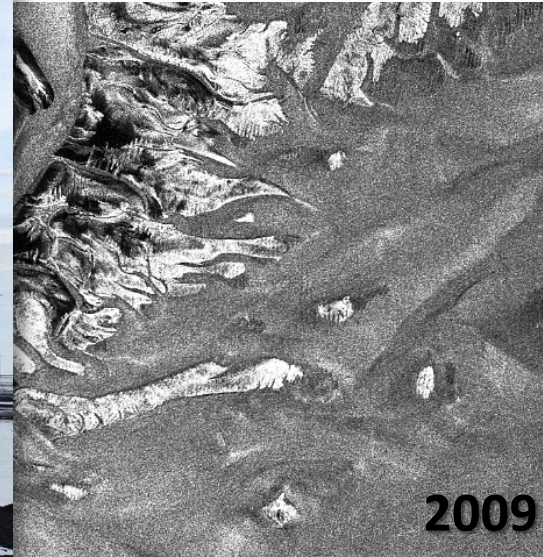


1993

ERS-1, 10 Jun 93, 0234 UTC



2011: Jiangsu Rudong intertidal offshore wind farm



2009

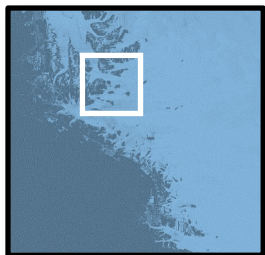
ENVISAT, 11 May 09, 1345 UTC



25 km x 25 km

2017

Sentinel-1A, 11 Mar 17, 0954 UTC



ReSCCoME: Remote Sensing of Changing Coastal Marine Environments

Full archive of wind maps from DTU:

<https://science.globalwindatlas.info/> (select 'Offshore wind fields in near-real-time')



ACTIVE LAYER: SAR WINDS (NEAR-REAL-TIME)

DATE FROM:	DATE TO:
YYYY-MM-DD	YYYY-MM-DD
LATITUDE FROM:	LATITUDE TO:
17.926673	36.115059
LONGITUDE FROM:	LONGITUDE TO:
102.311340	124.451165

TOTAL RECORDS: 8207 PAGE: 1 ENTRIES: 165 LOAD DATA: GO PAGE SIZE: 50

2021/06/19 - 09:48:00

2021/06/18 - 09:56:51

2021/06/18 - 09:56:16

2021/06/18 - 09:56:01

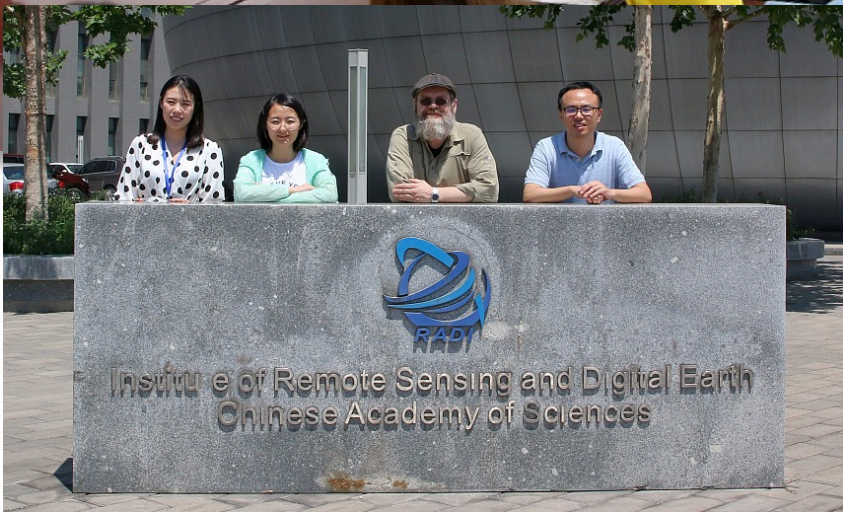
2021/06/18 - 09:55:36

2021/06/18 - 09:55:11

In-situ Data Measurements and Requirements:

- Intertidal Flats: surface roughness, sediments, moisture, habitats
- Offshore Wind Farms: atmospheric and oceanic parameters
- Offshore Oil Pollution: surface films, environmental conditions
- Coastal Pollution: plastic debris, waves & currents, bathymetry
- Coastline Changes: waterlines, water level, bathymetry

Training Classes at AIRCAS, Beijing, and TJU, Tianjin



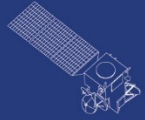
Field Data Collection Campaigns and Periods:

- UHH student excursions on the German North Sea coast





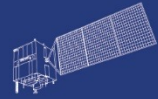
57192 - ReSCCoME



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CBERS



Gaofen



Beijing-2



Sentinel-1



Sentinel-2



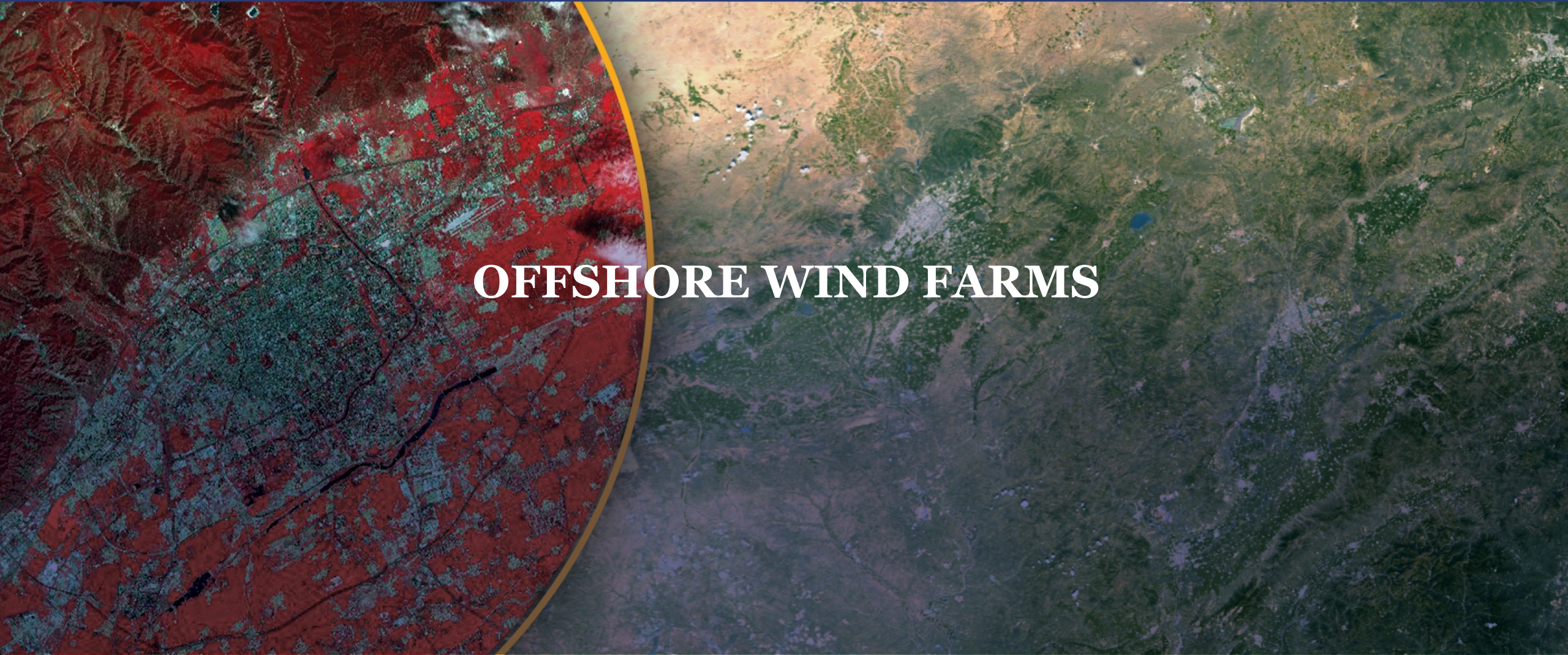
Sentinel-3



Sentinel-5p



Aeolus



OFFSHORE WIND FARMS

SAR Wind Retrieval and Wakes for Offshore Wind Energy Applications in East China

Credit: Abdalmenem Owda

Wind wakes:

- Defined as wind-speed reduction at downwind side of offshore wind farms (OWFs)
- Decrease efficiency of OWFs
 - reduce wind-power generation
 - increase load effects on wind turbines

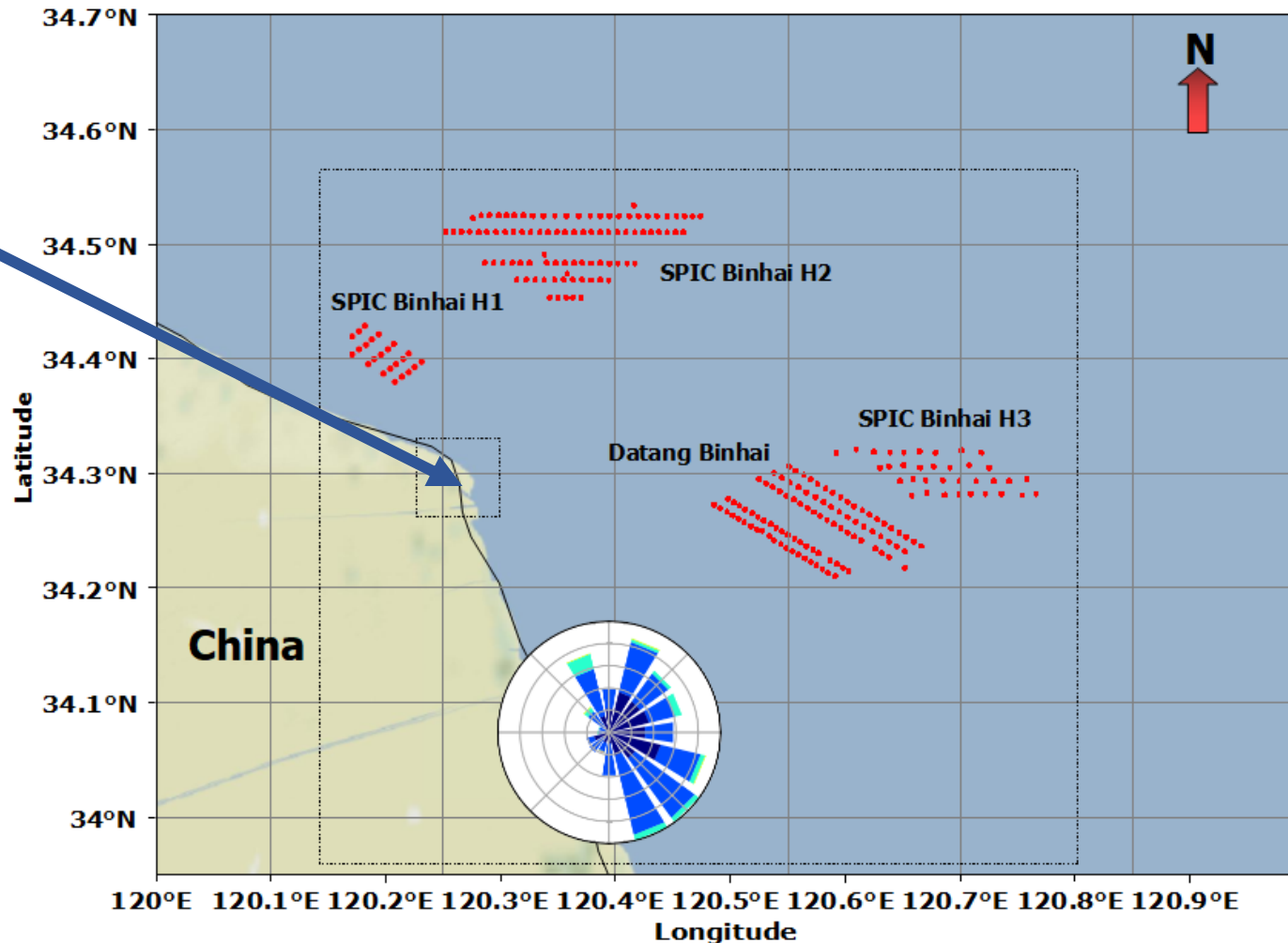
SAR:

- Provide wind speed (10 m) and wakes deficit at high spatial resolution (min. 500 m)
- Wind wakes cause high local radar contrast at downwind sides of OWFs

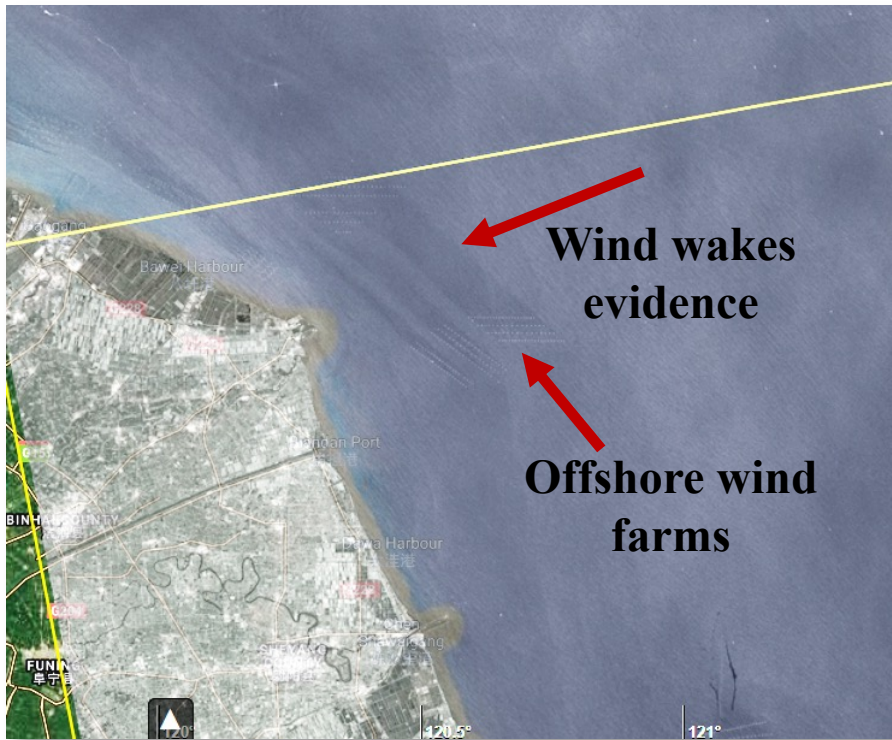
Study area on the Chinese coast

land expansion
into the sea

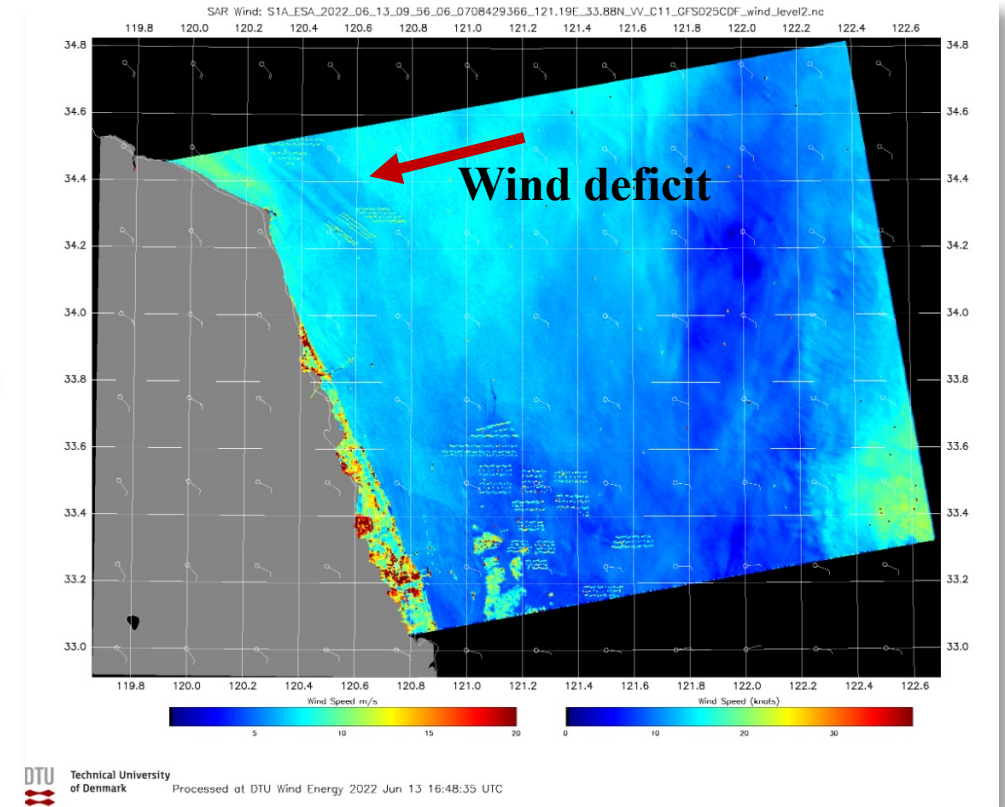
may increase
local std dev of
wind speed
derived from
SAR data



SAR Wind retrieval for offshore wind energy applications in E China



CMOD5.N

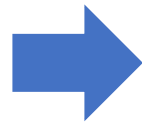
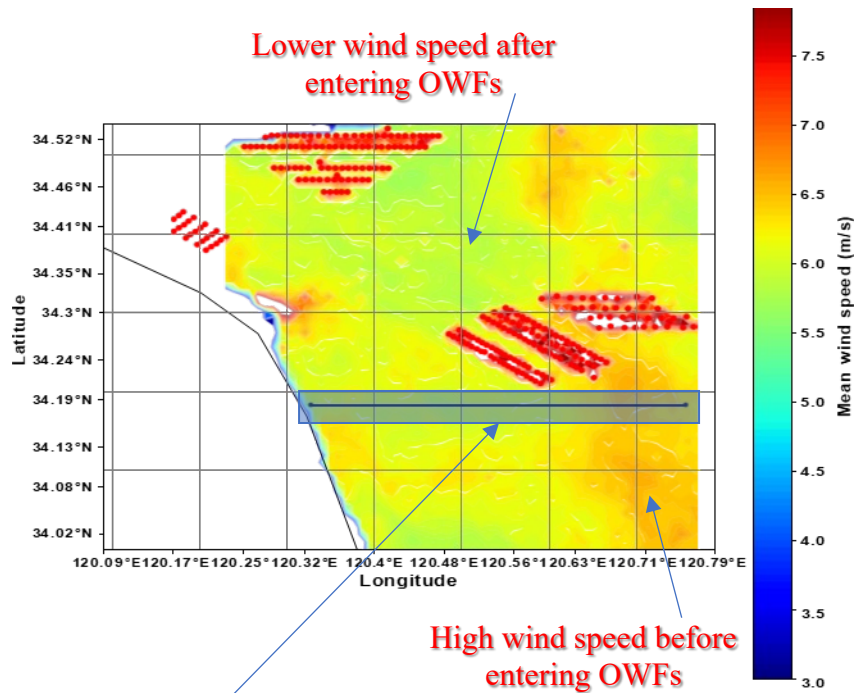


Sentinel 1A-level 1-GRDH was taken at 13^h
 June 2022 13:09:55
<https://ovl.oceandatalab.com/>

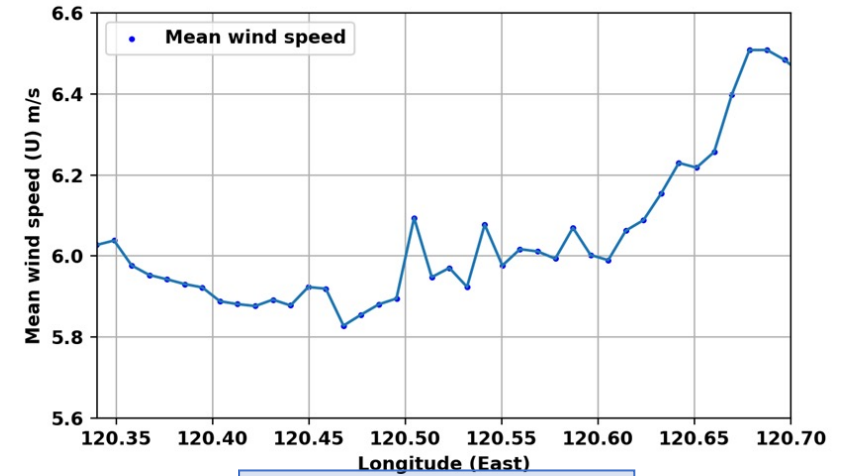
SAR wind map for the same Sentinel scene
<https://science.globalwindatlas.info/>

Wind speed and power variation near coast

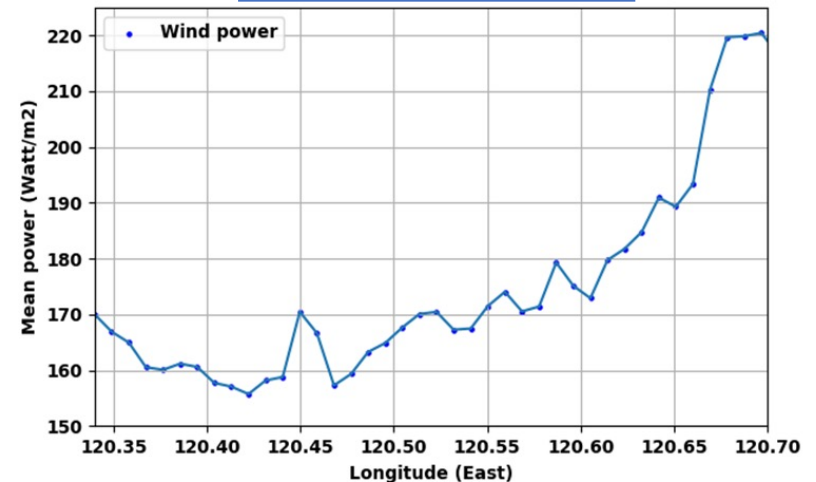
Average wind speed map from a group of SAR scenes



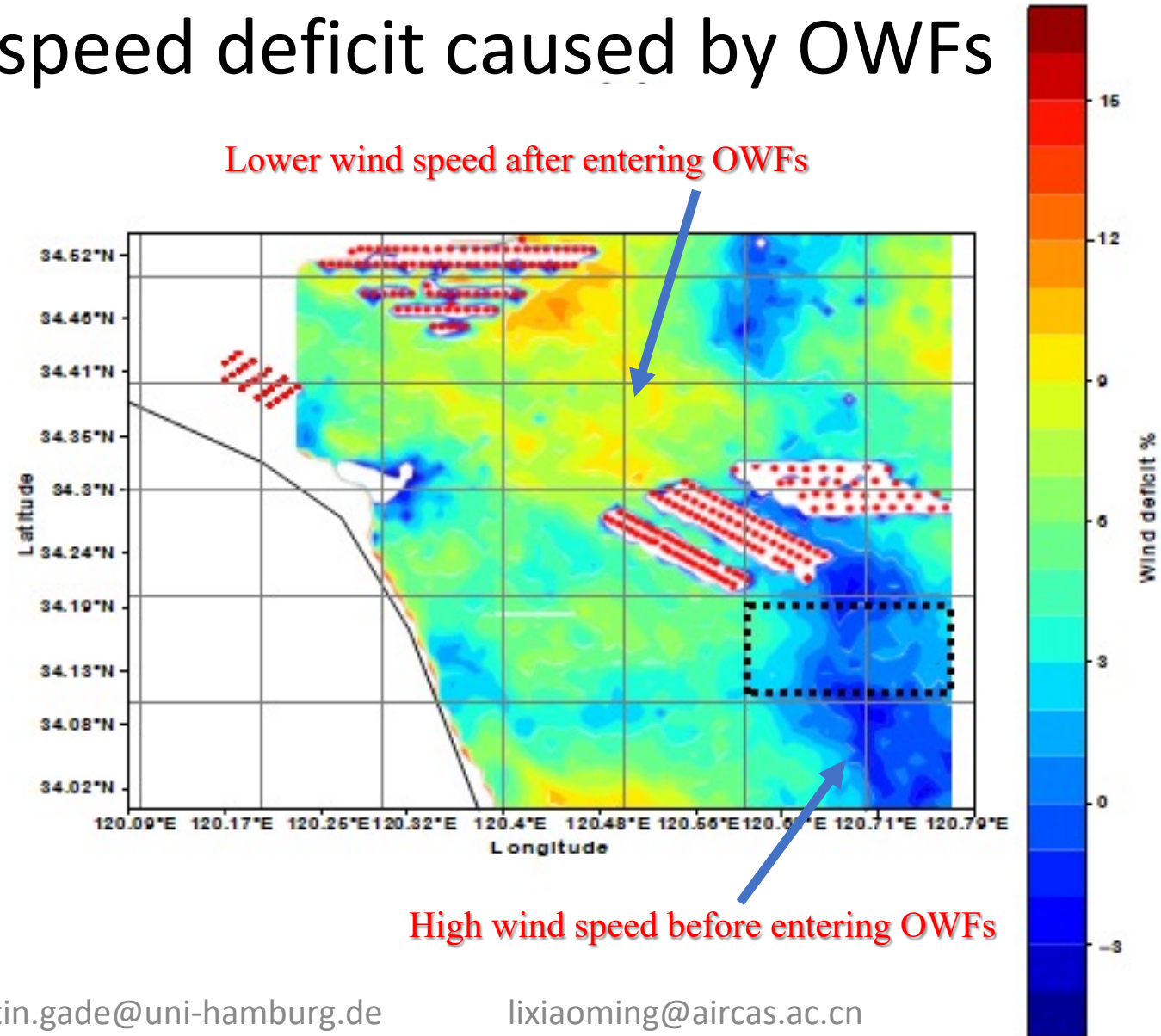
Wind speed and power variation along the black line (left). More wind speed reduction and power losses getting closer to the coast



$$P_{mean} = \frac{1}{2} \times \rho \times U^3$$

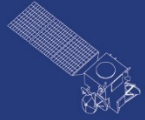


Wind speed deficit caused by OWFs





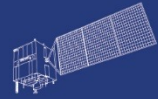
57192 - ReSCCoME



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Gaofen



Beijing-2



Sentinel-1



Sentinel-2



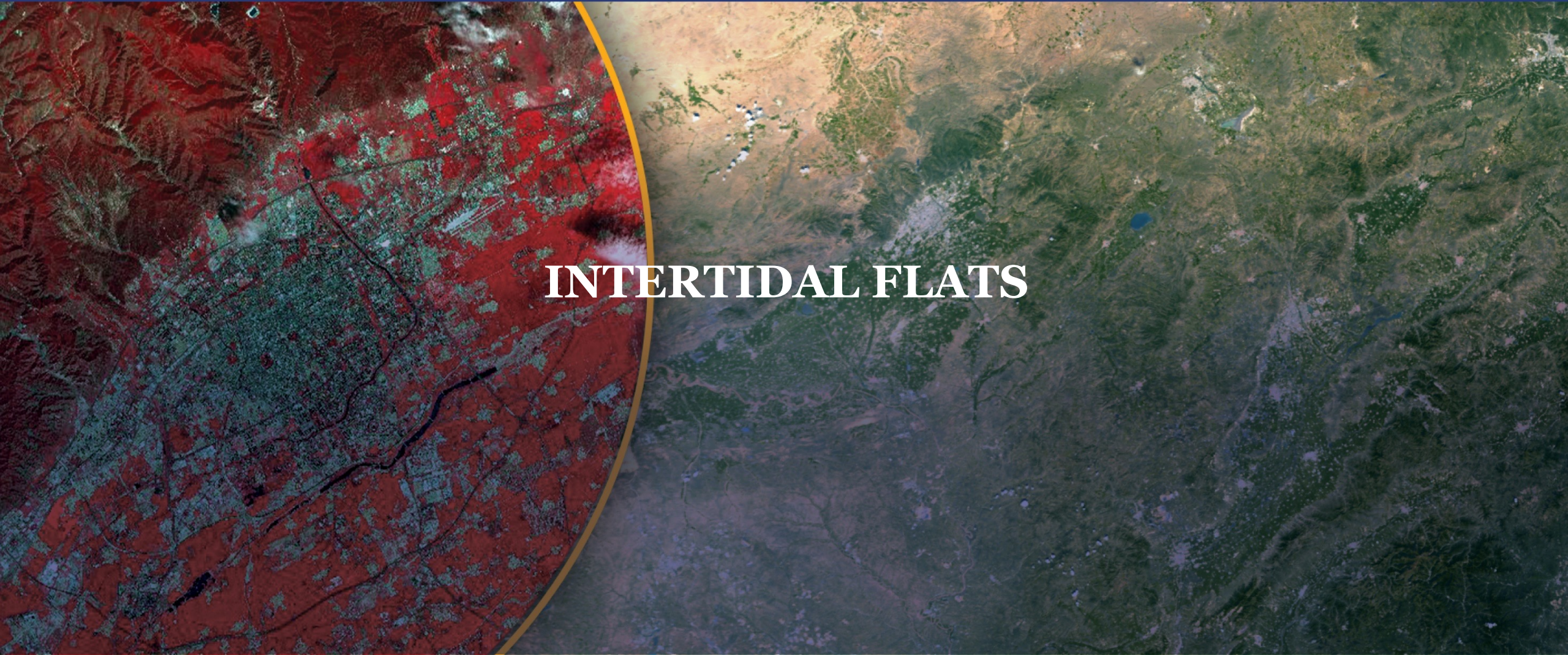
Sentinel-3



Sentinel-5p



Aeolus



INTERTIDAL FLATS

Classification of Intertidal Flat Surfaces by Means of Multi-Sensor Data

Credit: Zhang Di, Wang Wensheng

Wadden Sea: intertidal flats on the European North Sea coast

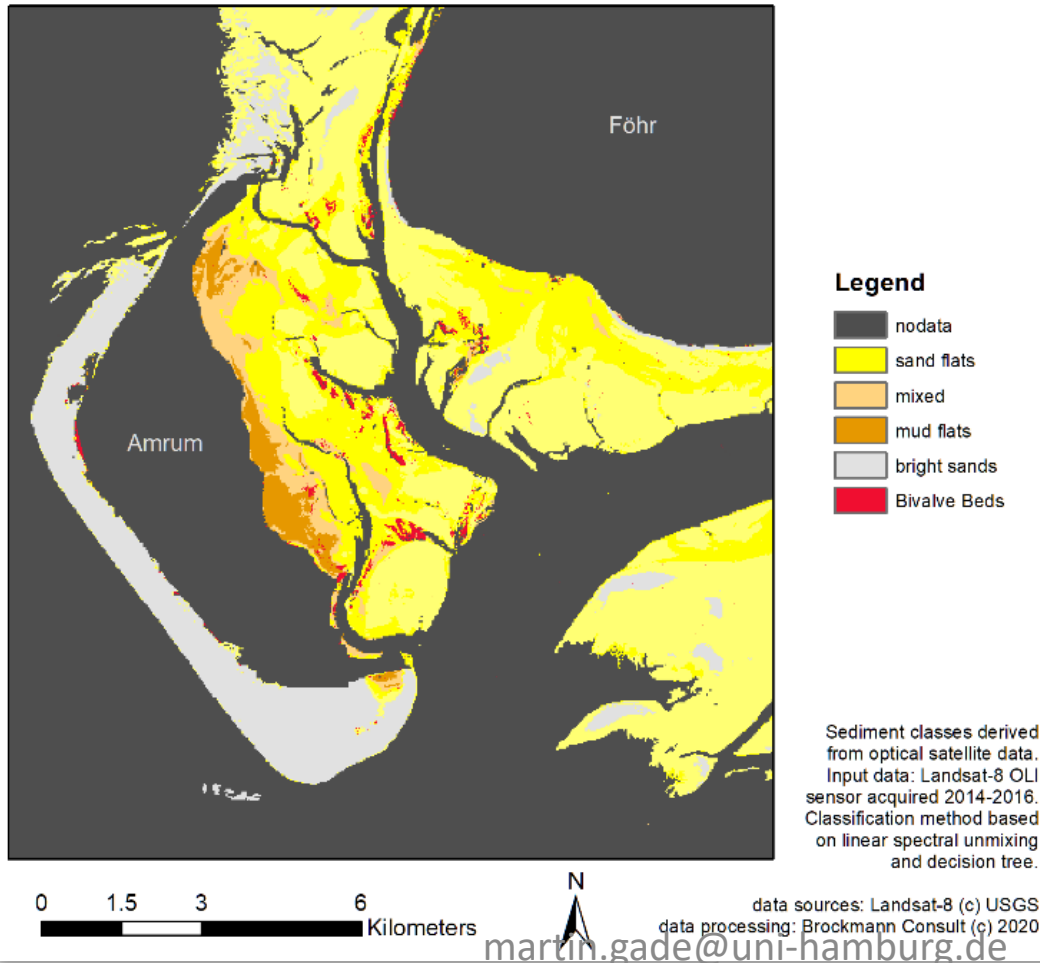


Intertidal flats:

- ~ 10 km offshore, exposed during low tide
- Danish-German-Dutch North Sea coast
- Fine sediments, partly vegetated, habitats
- Strong morphodynamics
- Invasive species: Pacific Oyster (*Crassostrea gigas*)
- National Parks, UNESCO World Heritage (2009)
- Frequent monitoring mandatory
- SAR, because of usually high cloud cover

Traditional classification based on optical data

Sediment classes derived from Optical Satellite data



5 surface classes:

- sand
- mixed sand/mud
- mud
- bivalves
- bright sand

- Linear spectral unmixing
- Decision tree

Data source:

Landsat-8

Classification:

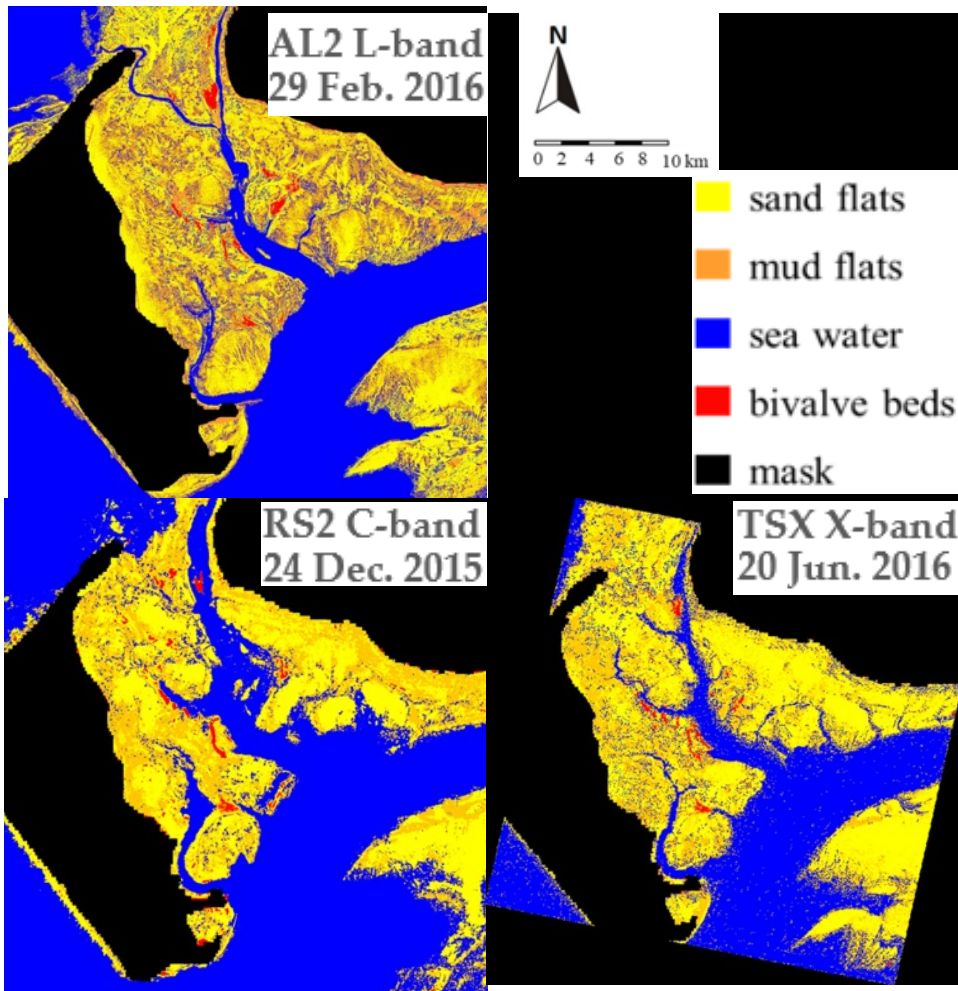
© Brockmann Consult

lixiaoming@aircas.ac.cn



6 km × 6 km

Discrimination of Surface Types — FCDK-RF Algorithm



FCDK-RF

- Freeman-Durden:
 - FD_odd, FD_vol, FD_dbl;
- Cloude-Pottier:
 - Entropy H, alpha angle α , Anisotropy A;
- Double-Bounce Eigenvalue Relative Difference (DERD);
- Kennaugh Elements (D3 and P)
- Classification according to Random Forest (RF) theory.

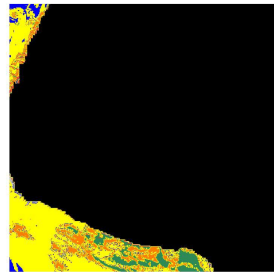
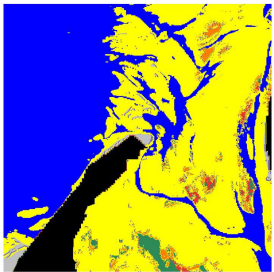
EXTENSION

- Partly Polarimetric Mode (HH with VV);
- Sediments, Habitats, Bivalve (oysters and mussels) beds.

	AL2 (L-Band)		RS2 (C-Band)		TSX (X-Band)	
	PA (%)	UA (%)	PA (%)	UA (%)	PA (%)	UA (%)
Sandy sediments	86.7	85.6	89.1	86.2	81.7	79.8
Mixed sediments	86.8	87.2	88.0	86.7	81.0	80.9
Open water	86.5	88.5	88.9	87.6	86.2	87.0
Bivalve beds	89.0	91.8	90.9	91.3	91.6	92.7
OA (%)	86.2		88.7		85.9	

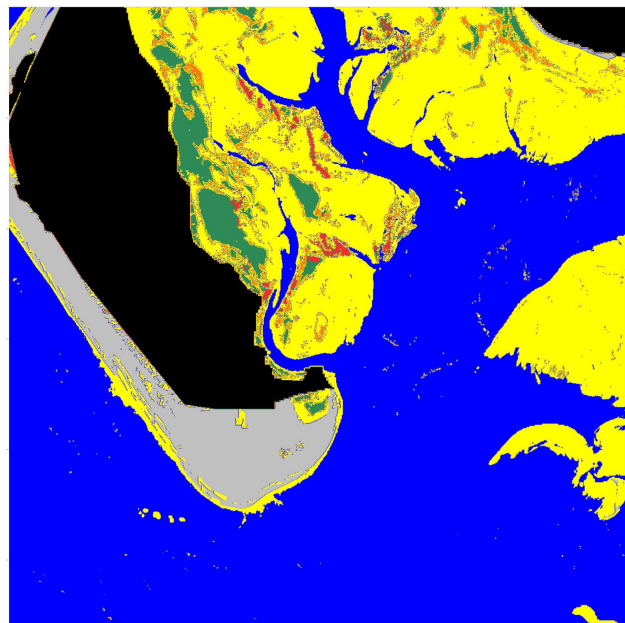
TE-UNet: Texture Enhancement UNet

■ Land
 ■ Seagrass
 ■ Bivalves
 ■ Beach
 ■ Water
 ■ Sediments
 ■ Thin Coverage

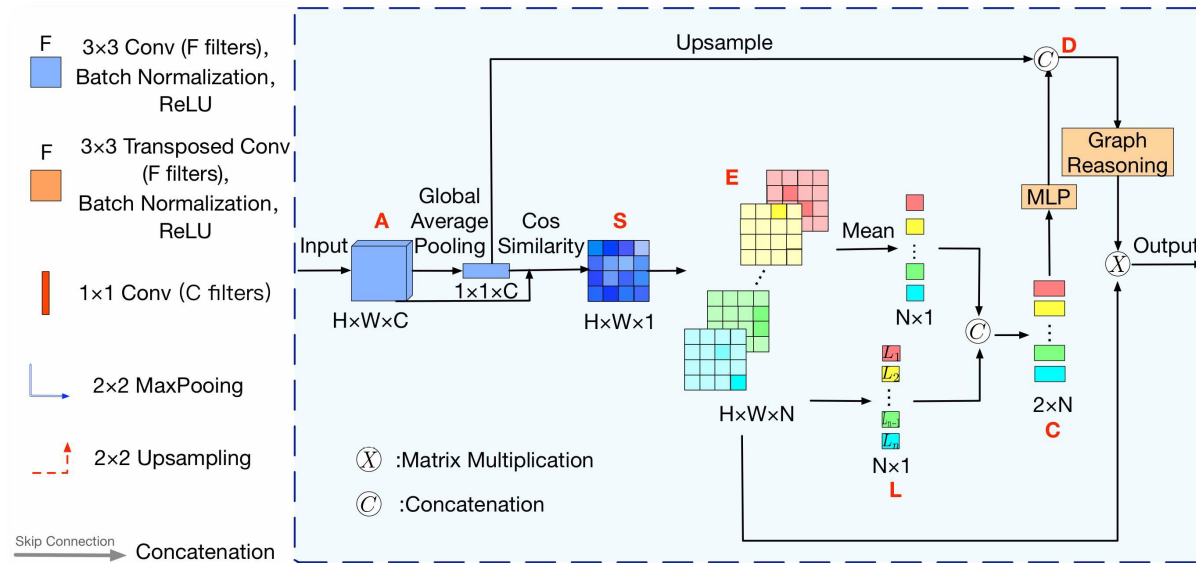


(a) Testing Site 1 (512 × 512)

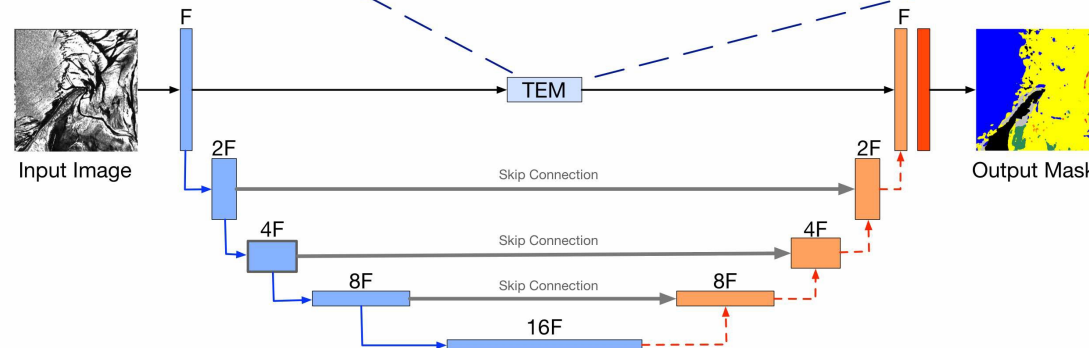
(b) Testing Site 2 (512 × 512)



(c) Training Site (1187 × 1187)



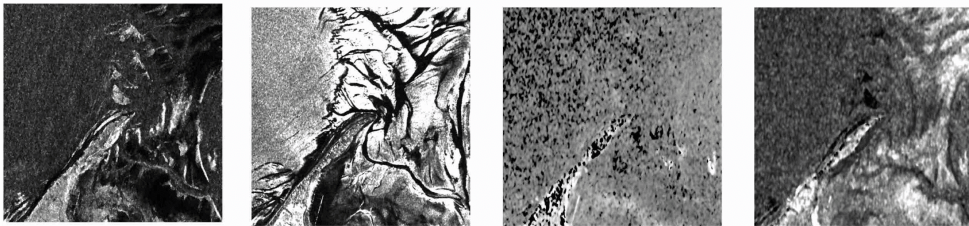
6 km × 6 km



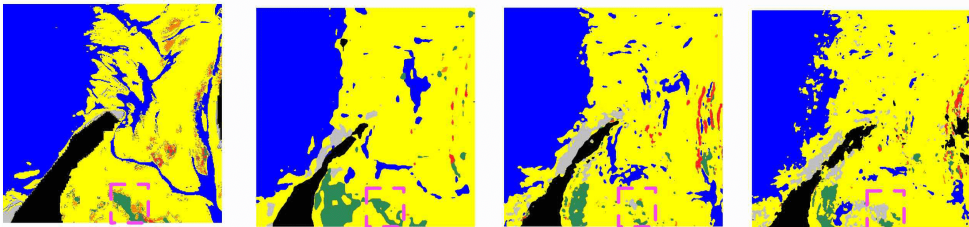
[Zhang 2022]

Classification Based on Deep Learning Using SAR and Optical Data

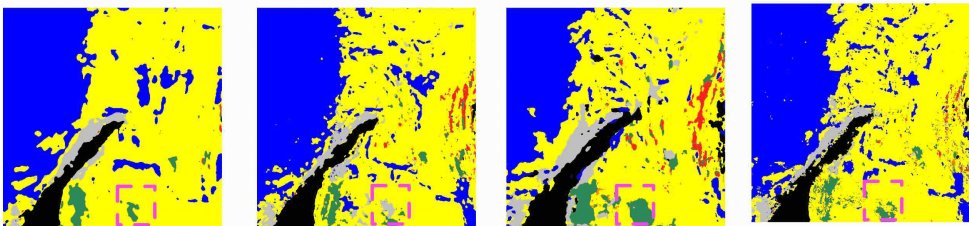
■ Land
 ■ Seagrass
 ■ Bivalves
 ■ Beach
 ■ Water
 ■ Sediments
 ■ Thin Coverage



SAR Images (RS2 VV Channel) SAR Images (ALOS2 VV Channel) SAR Images (RS2_FD_dbl_r) SAR Images (RS2_CP_Alpha)



Reference Classification CPI (Proposed) I FD



CP FDI FDCP FDCPI

Radarsat-2
 24 Dec 2015, 0543 UTC

ALOS-2
 29 Feb 2016, 2257 UTC

Reference:

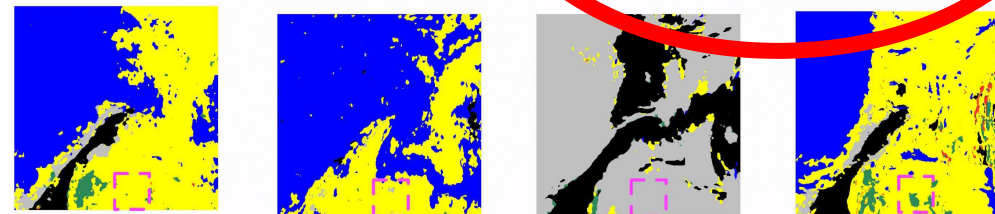
Landsat-8, SPOT-4, © Brockmann Consult



6 km × 6 km



SAR Images (RS2 VV Channel) SAR Images (ALOS2 VV Channel) Reference Classification Training: RS2+ALOS2 Testing: RS2+ALOS2 (Proposed)



Training: RS2 Testing: RS2 Training: ALOS2 Testing: RS2 Training: RS2 Testing: ALOS2 Training: ALOS2 Testing: ALOS2

[Zhang 2022]

On the Use of SAR Data to Monitor Coastal Erosion and Morphodynamics in Intertidal Areas

Credit: Sebastian Peters, Simon Schäfers

Coastal erosion caused by storm surges



Norderney 2022

martin.gade@uni-hamburg.de

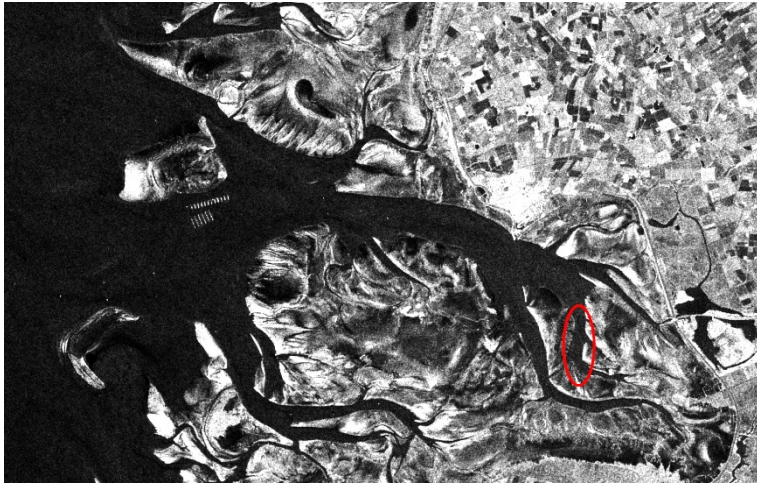


lixiaoming@aircas.ac.cn

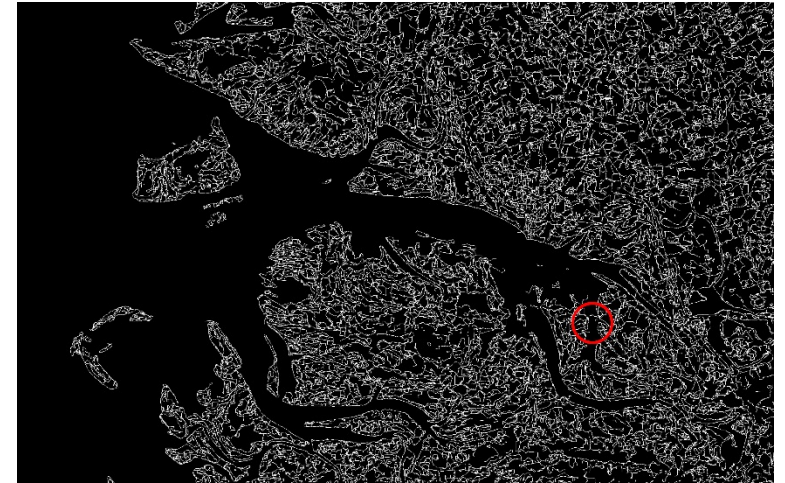


Langeoog 2022 28

Wadden Sea waterlines from Sentinel-1 SAR data



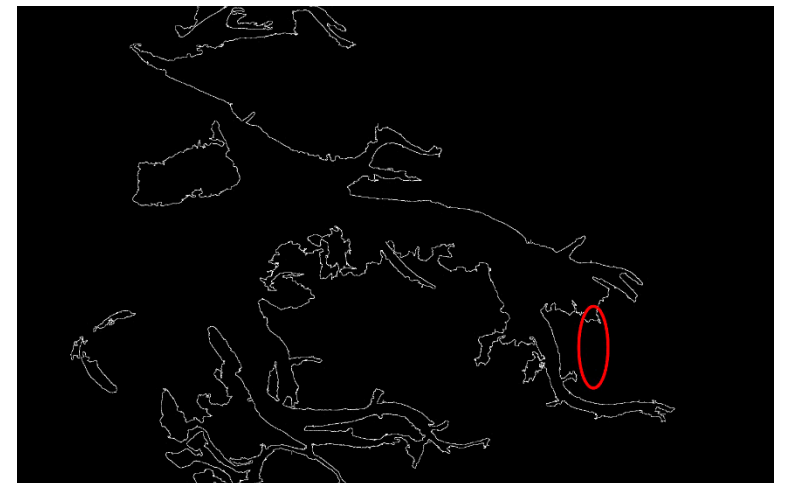
original SAR image



detected edges

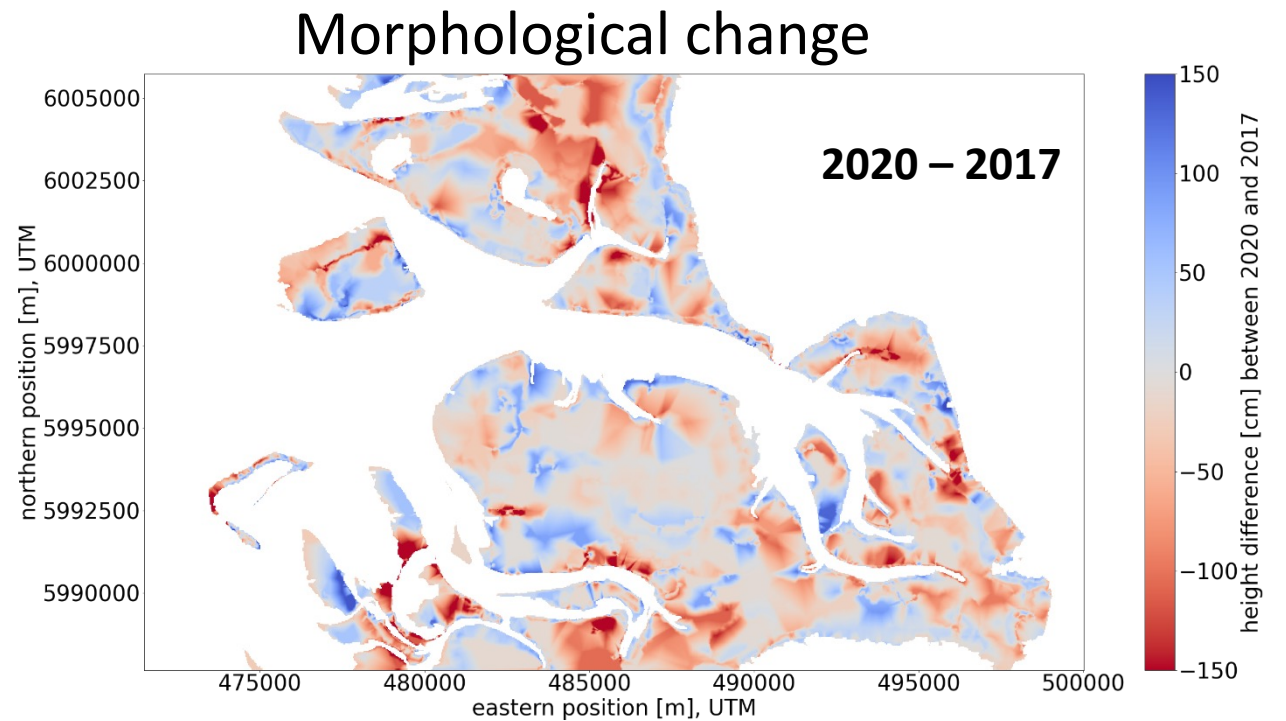
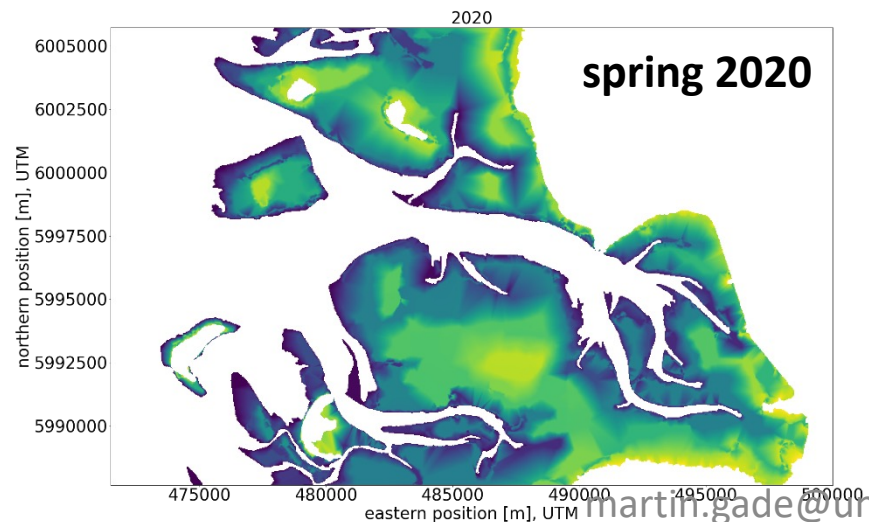
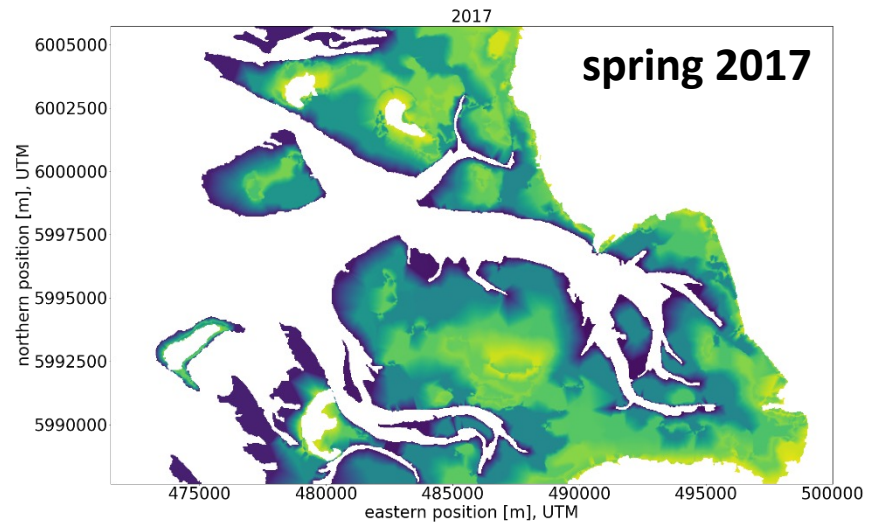


binary land-water map



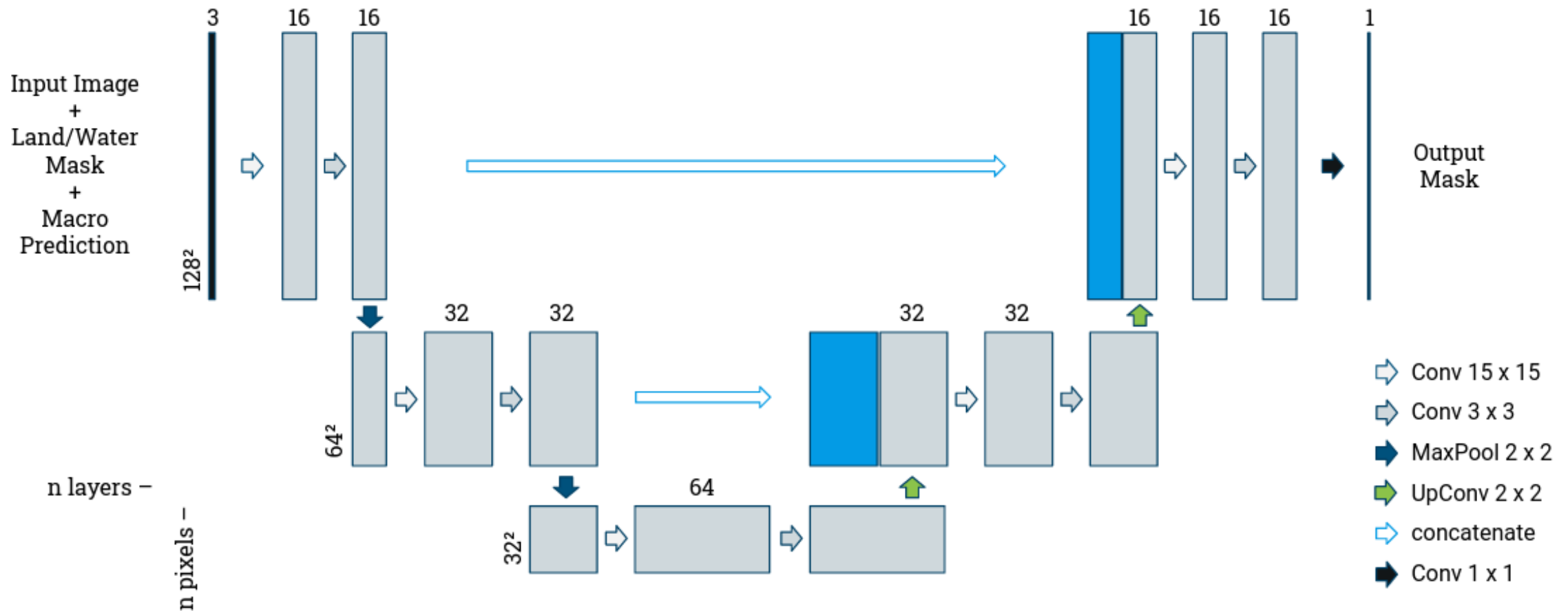
final waterlines

Wadden Sea topography from waterlines



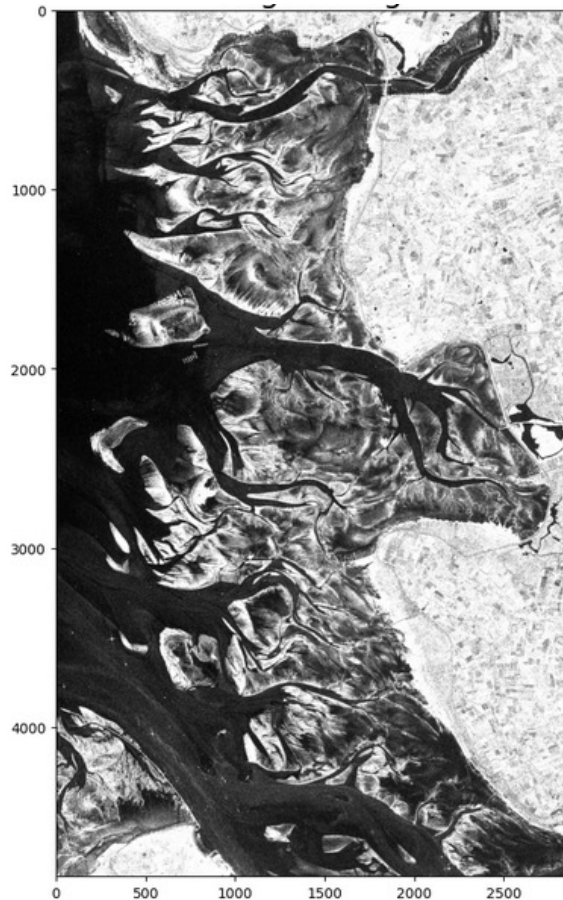
Classification aided by a neural network

Convolutional Neural Network (CNN) : U-Net

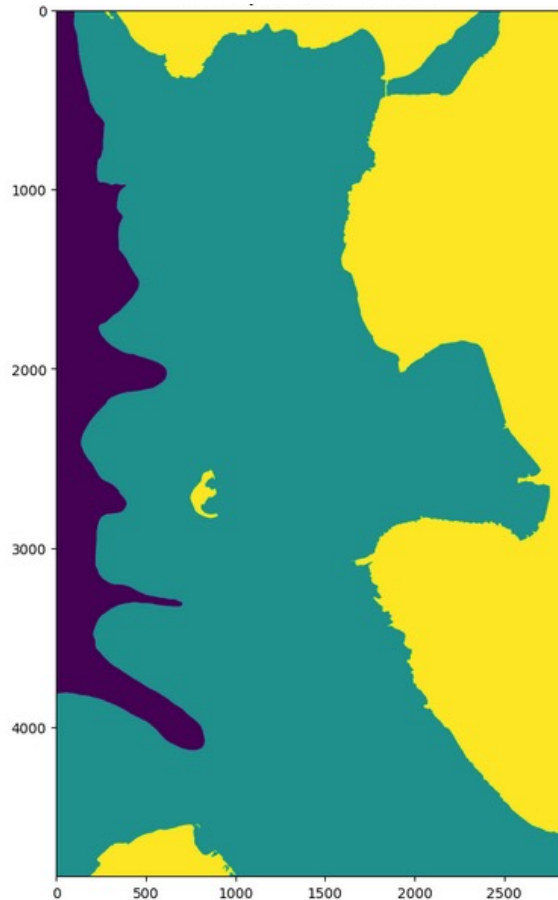


Classification aided by a neural network

Sentinel-1A SAR image
19 May 2020

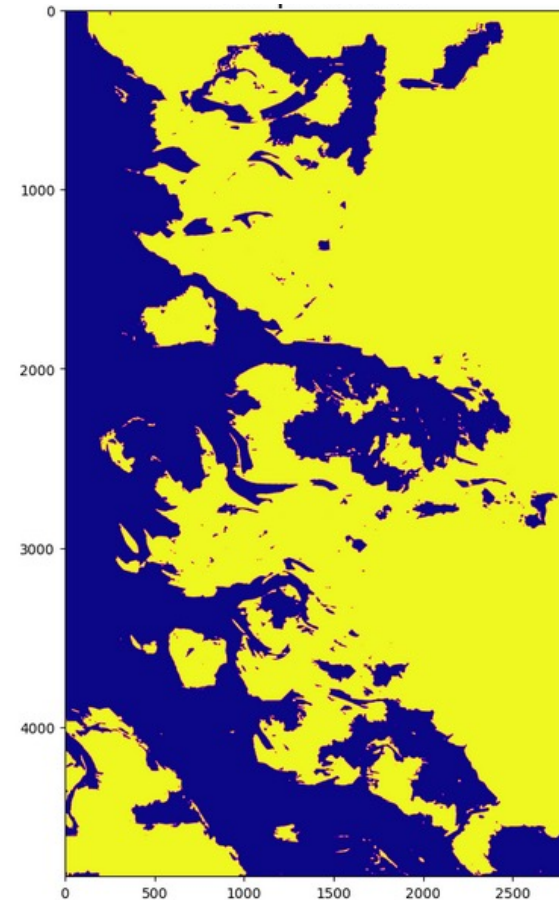


additional input
water – **transition** – **land**



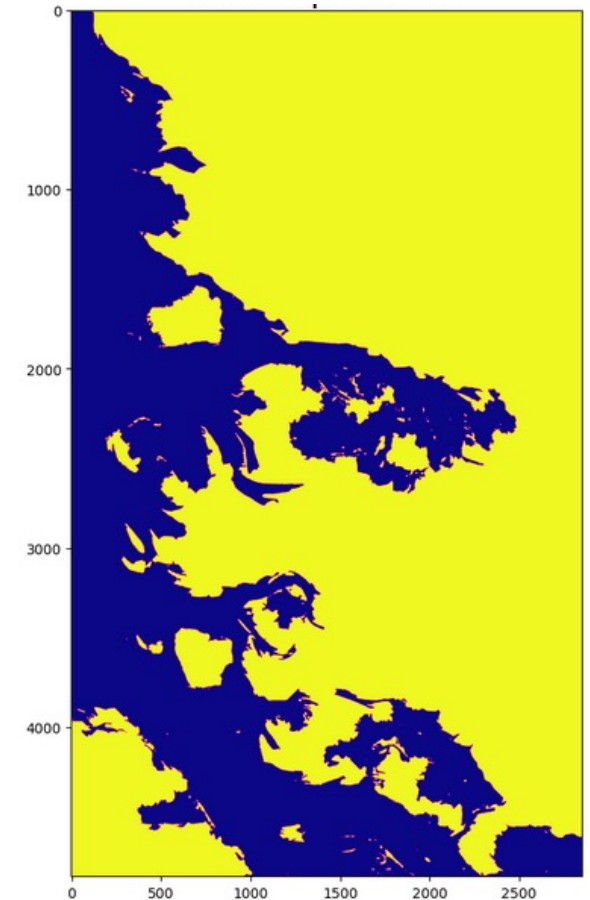
martin.gade@uni-hamburg.de

classification output
land / water mask



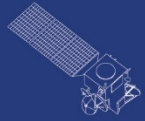
lixiaoming@aircas.ac.cn

refined output
after floodfilling





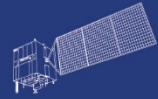
57192 - ReSCCoME



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Gaofen



Beijing-2



Sentinel-1



Sentinel-2



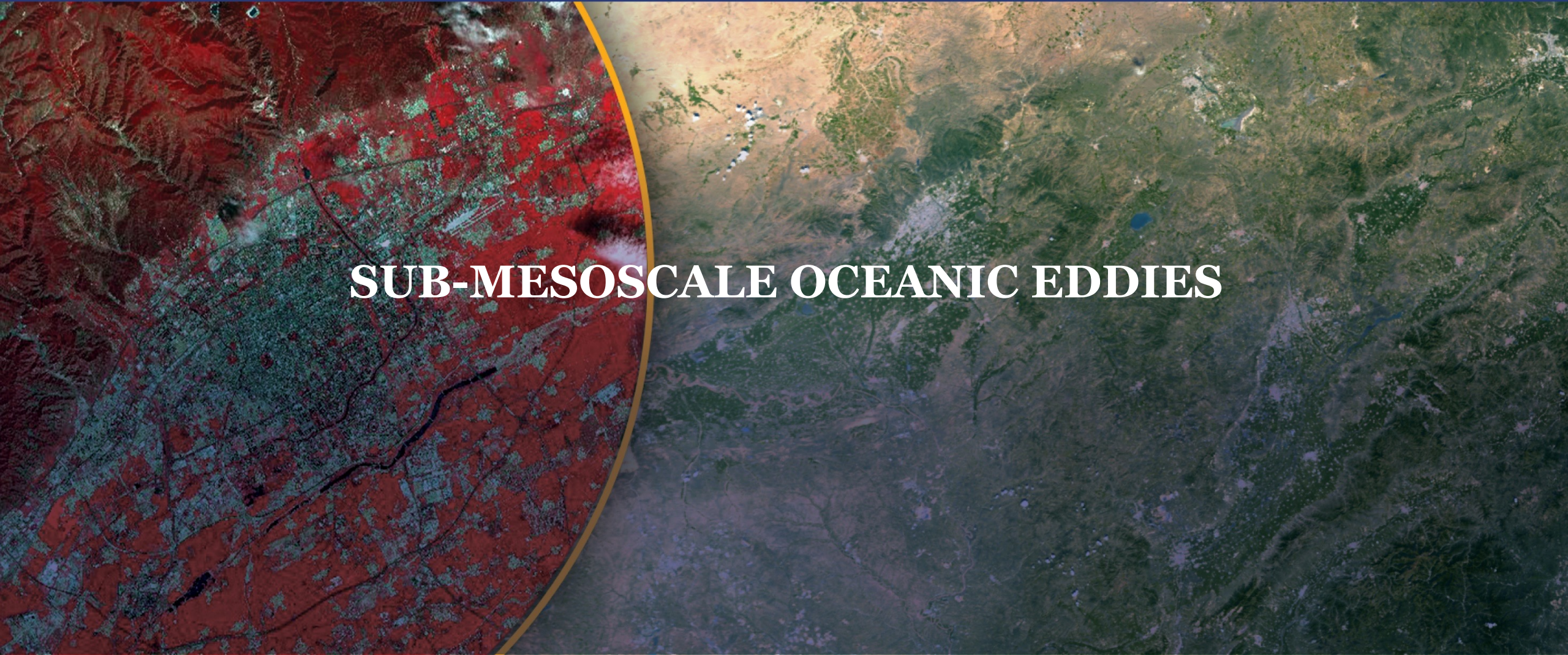
Sentinel-3



Sentinel-5p



Aeolus



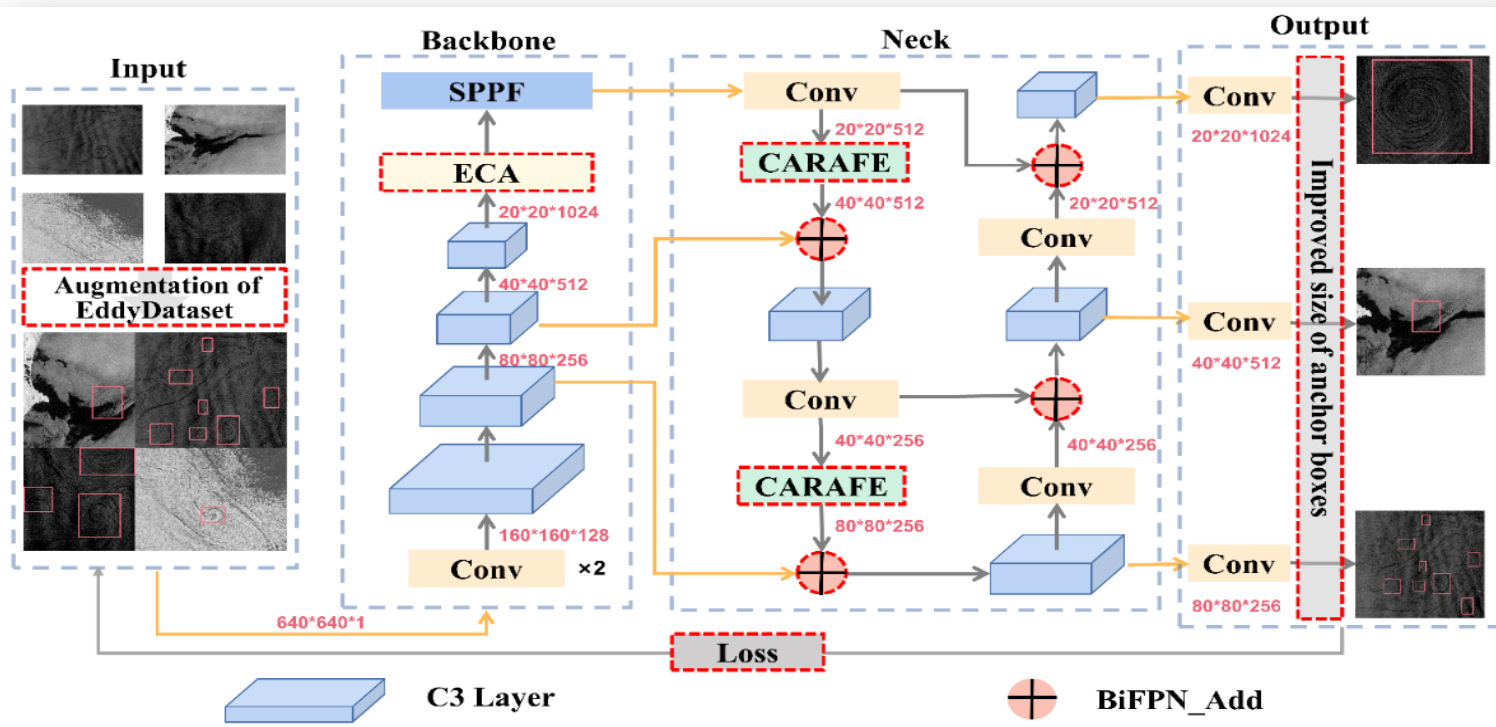
SUB-MESOSCALE OCEANIC EDDIES

Oceanic Eddy Detection from SAR Imagery Based on Deep Learning Network

Credit: Zi Nannan

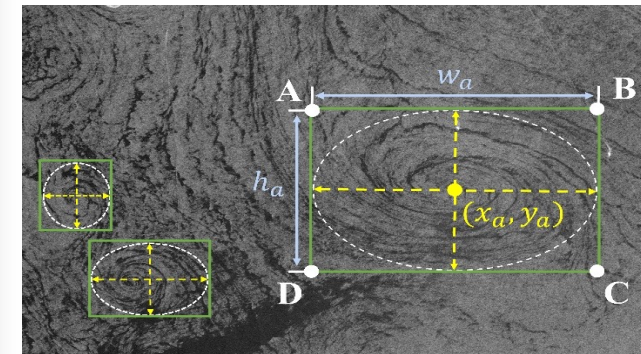
Overall architecture of EOLO

◆ The automatic detection network of EOLO



Eddy automatic detection network structure based on YOLO model (EOLO)

◆ Geographic information extraction of detected eddies



Eddy center: (x_a, y_a)
 Diameter: $\frac{AB + AD}{2}$

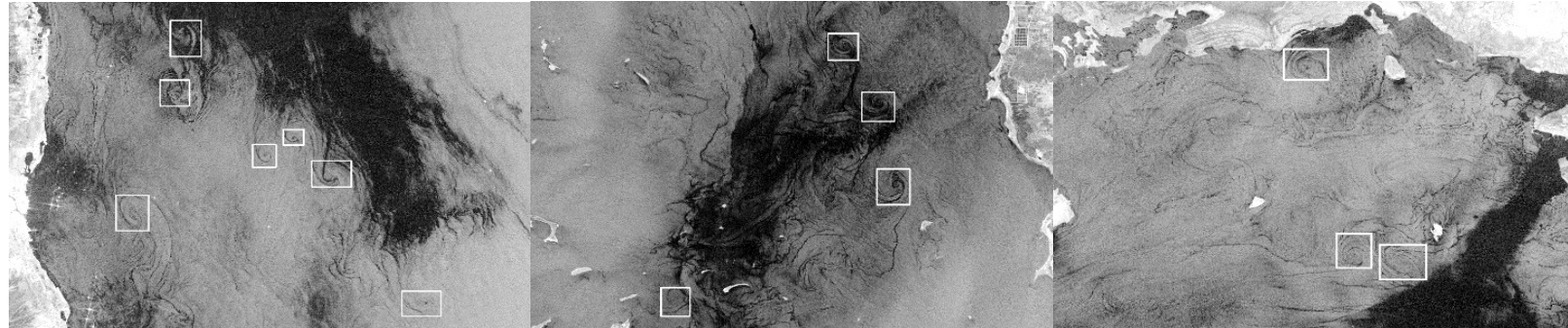
Geographic Affine Transformation model

Convert to row and column numbers to the corresponding geographic coordinates $(x_a, y_a, w_a, h_a) \rightarrow$ **longitude and latitude of eddy center and eddy diameter**

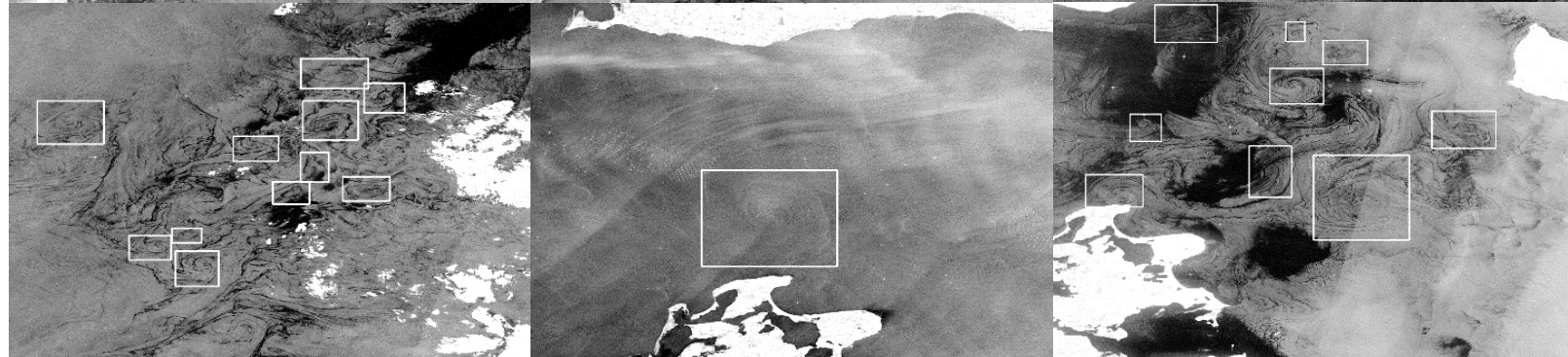
EOLO outputs detected eddies with their center and width-height information

Adaptability of EOLO in different marginal seas

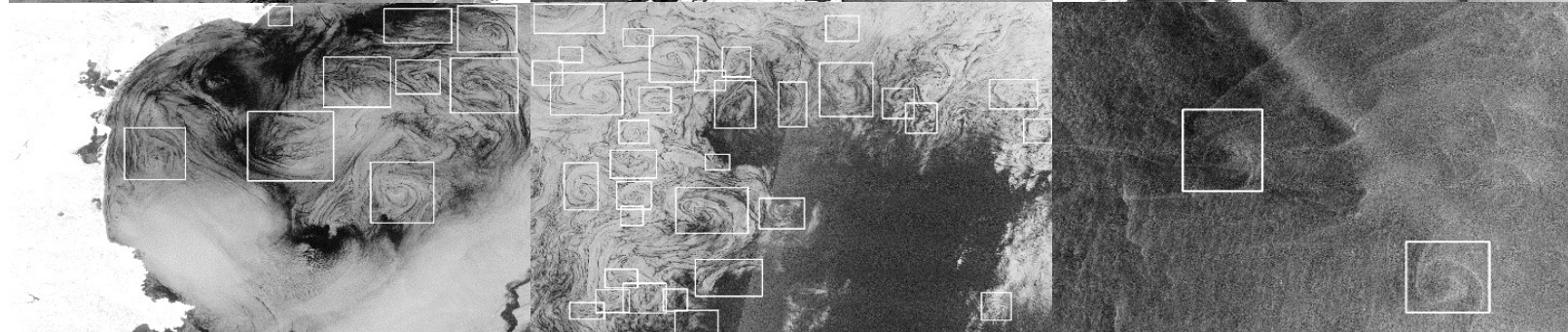
Red Sea
(Precision = 0.968)



Baltic Sea
(Precision = 0.957)



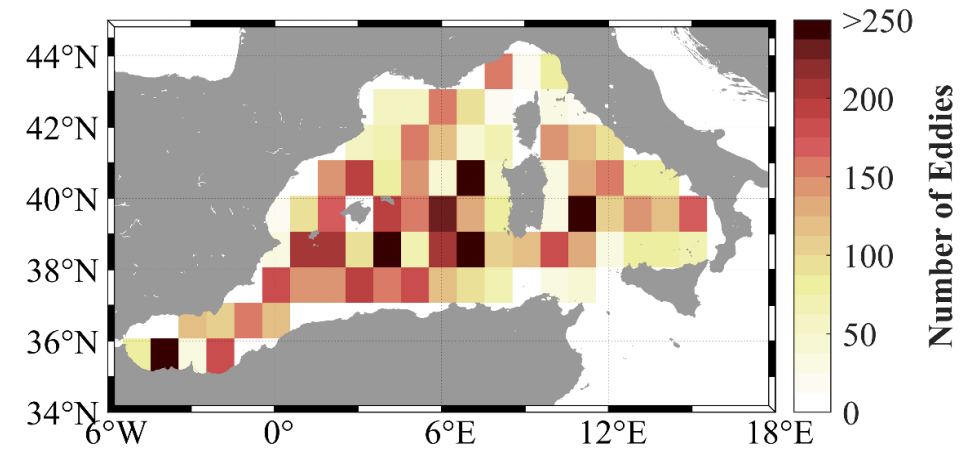
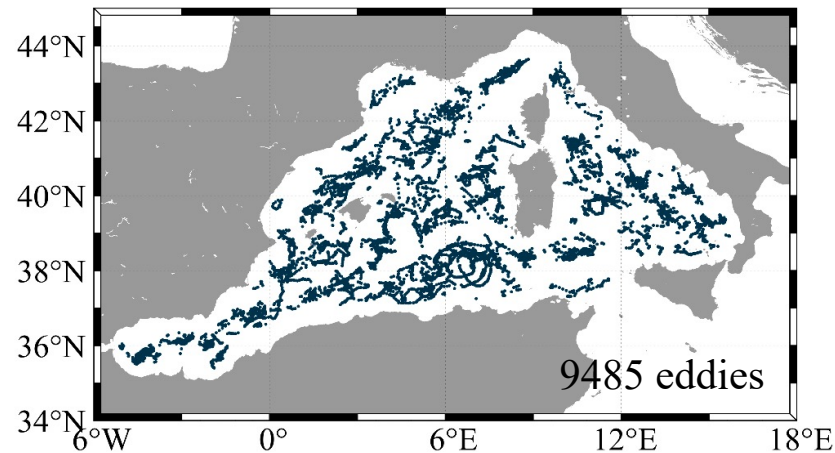
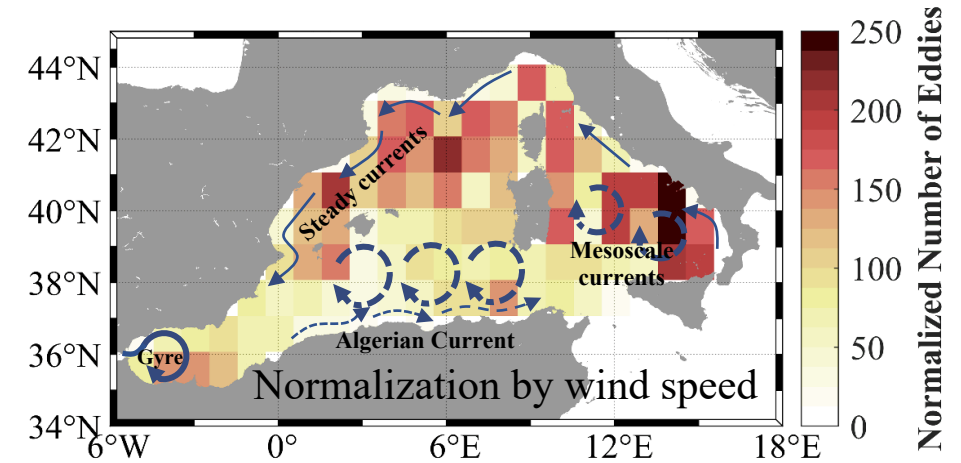
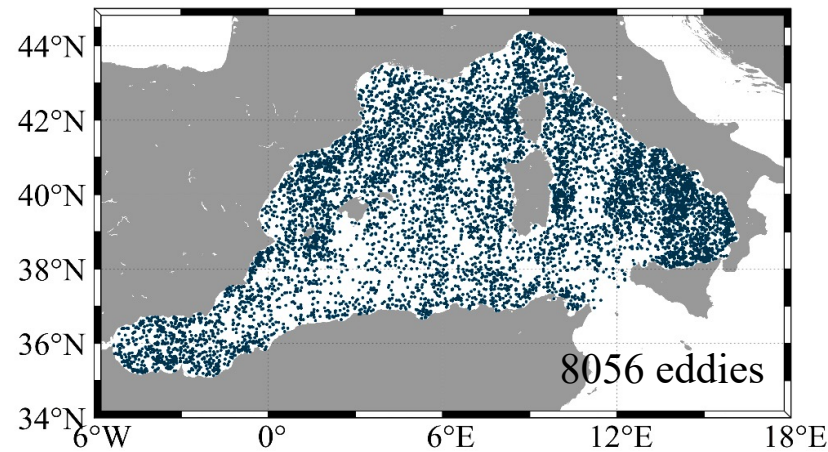
**Western
Mediterranean Sea**
(Precision = 0.965)



Spatial characteristics of ocean eddies in the Western Med

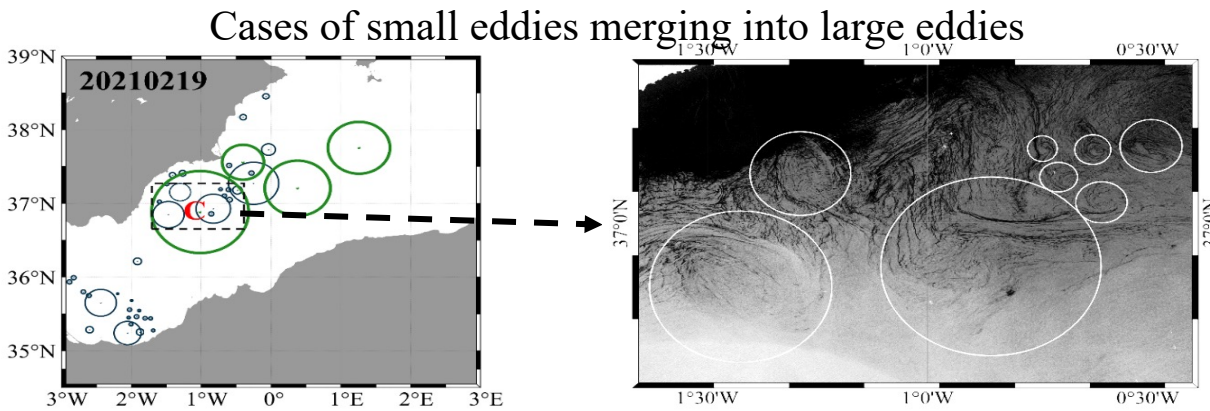
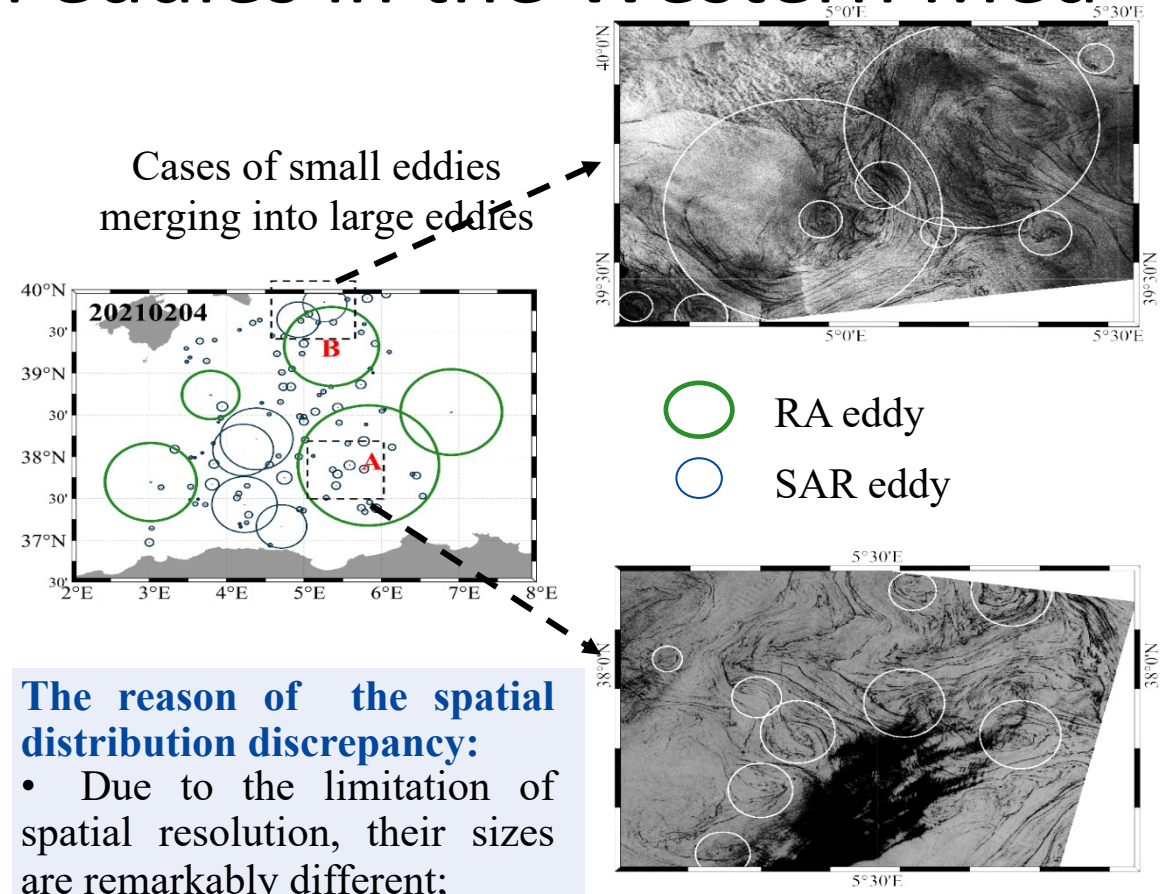
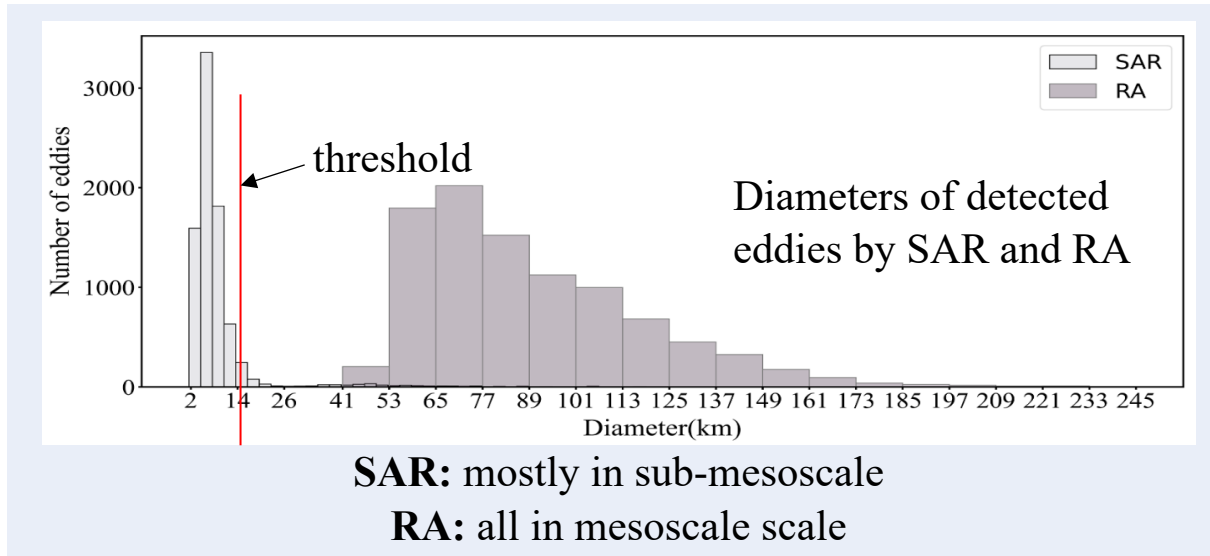
Data Source
 In 2021
SAR:
 Sentinel-1 A/B

RA:
 Global altimeter eddy
 product provided by
 AVISO (META3.2 DT)



Notable differences in the spatial distributions of ocean eddies detected by SAR and RA.

Spatial characteristics of ocean eddies in the Western Med



The reason of the spatial distribution discrepancy:

- Due to the limitation of spatial resolution, their sizes are remarkably different;
- Also due to the limitation of spatial resolution, the eddies detected by gridded map merging of multi-RA along-track data may be the results of small-scale ones merging into larger ones.

- Gade, M., M. Badger, K. Dimitriadou, X. Li, A. Owda, S. Peters, S. Schäfers, D. Zhang, 2023: “Remote Sensing of Changing Coastal Marine Environments – a Midterm Report”, *Trans. Geospatial. Inform. Serv.*, to be resubmitted.
- Zi, N., X.-M. Li, M. Gade, H. Fu, 2023: “Ocean Eddy detection based on YOLO deep learning algorithm by Synthetic Aperture Radar data”, *Remote Sens. Environ.*, under review.
- Zhang, D., M. Gade, W. Wang, H. Zhou, 2023: “EddyDet: A Deep Framework for Oceanic Eddy Detection in Synthetic Aperture Radar Images”, *MDPI Remote Sens*, revised.
- Zhang, D., W. Wang, M. Gade, 2023: “TENet: A Texture-Enhanced Network for Intertidal Sediments and Habitats Classification in Multi-band PolSAR Images”, *IEEE Trans. Geosci. Remote Sens.*, under review.
- Gade, M., S. Peters, S. Schäfers, 2023: “On the Use of SAR Data to Monitor Coastal Erosion and Morphodynamics in Intertidal Areas”, *MDPI Coasts*, in preparation.

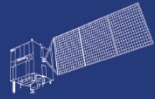
- Continuation of presented work
- Intensify field work, where needed
- Initiate new / foster existing collaborations
- Summer School together with partner project



HY



HJ-1AB



CBERS



Gaofen



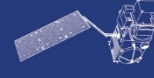
Beijing-2



Sentinel-1



Sentinel-2



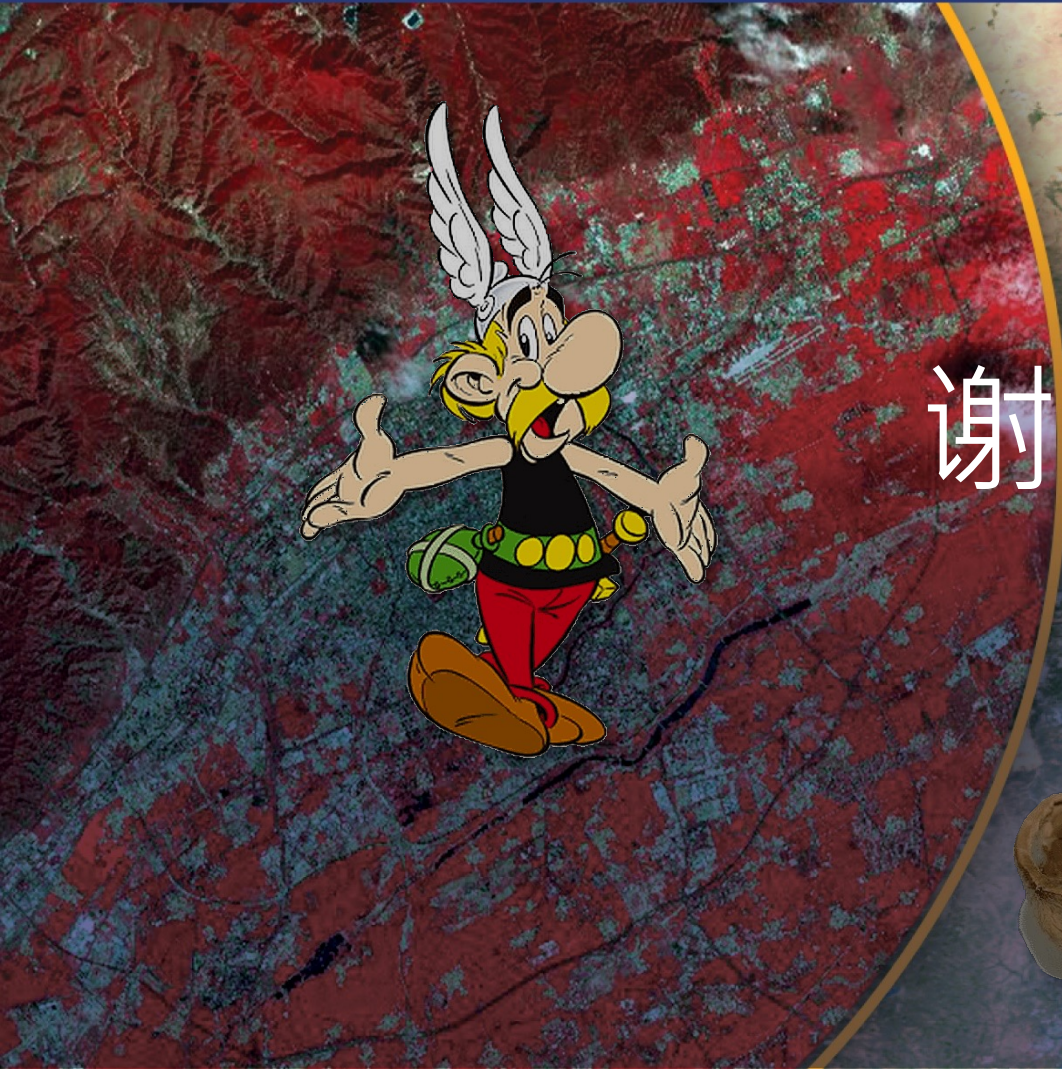
Sentinel-3



Sentinel-5p



Aeolus



谢谢！

