

# QuikSCAT-derived coastal winds



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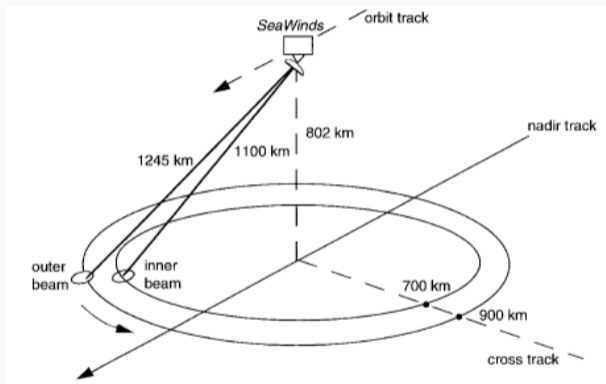
# Why coastal winds

- Scientific applications:
  - Weather forecasts (ocean and atmosphere)
  - Coastal dynamics (Diffusion, extreme events such as Acqua Alta, etc.)
  - ...
- Civil applications:
  - Wind farm installation
  - Coastal erosion
  - ...

**Scatterometer-derived winds represent the gold standard, but...**

**Land contamination causing wind biases within  $\approx 25$  km**

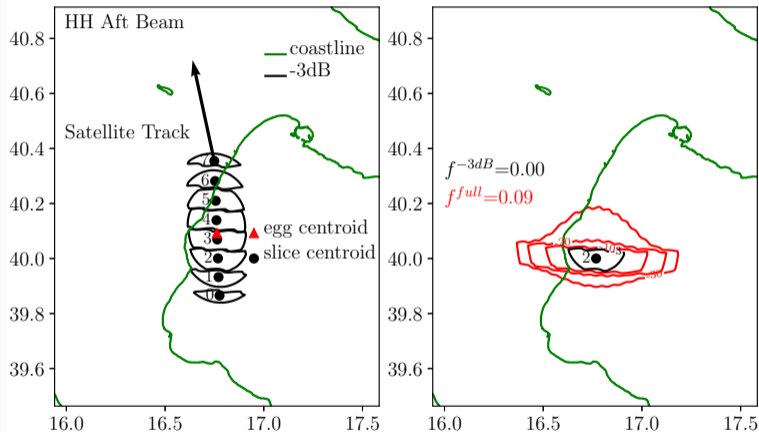
# SeaWinds on QuikSCAT and ADEOSII



- Ku-band (13.4 GHz)
- Inner beam: H-Pol
- Outer beam: V-Pol
- 4 "views" per each ocean point:
  - H-fore
  - H-aft
  - V-fore
  - V-aft

M. W. Spencer, Chialin Wu and D. G. Long, "Tradeoffs in the design of a spaceborne scanning pencil beam scatterometer: application to SeaWinds," in IEEE Transactions on Geoscience and Remote Sensing, vol. 35, no. 1, pp. 115-126, Jan. 1997, doi: 10.1109/36.551940.

# SeaWinds egg (Inner beam example)



- range filtering
- 8 slices per egg
- slice dimension  $\approx 25 \times 8 \text{ km}^2$
- $\forall$  slice, computation of **Land Contribution Ratio ( $f$ )**:

$$f = \frac{\sum_{xy} L_{xy} S_{xy}}{\sum_{xy} S_{xy}}$$

- $L_{xy}$ : land/sea mask
- $S_{xy}$ : Spatial Response Function (SRF)



# Model of land-contaminated $\sigma_0$

$$\sigma_0 = (1 - f) \bar{\sigma}_{0,S} + f \bar{\sigma}_{0,L} + \underbrace{[(1 - f) \epsilon_S(\sigma_{0,S}) + f \epsilon_L(\sigma_{0,L})]}_{\epsilon}$$

- $\bar{\sigma}_{0,S}$ ,  $\bar{\sigma}_{0,L}$ ,  $\epsilon_S$  and  $\epsilon_L$  are unknown

## State of the art

- $\bar{\sigma}_{0,L}$  estimated from the SeaWinds climatological series [1] or enhanced res.algorithm [2]
- $\hat{\bar{\sigma}}_{0,S} = \frac{\sigma_0 - f \bar{\sigma}_{0,L}}{1 - f}$

[1] Fore, A.G.; Stiles, B.W.; Strub, P.T.; West, R.D. QuikSCAT Climatological Data Record: Land Contamination Flagging and Correction. Remote Sens. 2022, 14, 2487. <https://doi.org/10.3390/rs14102487>

[2] Soisuvarn, S., Jelenak, Z., Chang, P.S., Zhu, Q., Shoup, C.G., 2023. High-resolution coastal winds from the noaa near real-time ascat processor. IEEE Transactions on Geoscience and Remote Sensing 61, 1–12. doi:10.1109/TGRS.2023.3279764.

# $f$ -based $\sigma_0$ correction scheme with noise regularization

Noise regularization procedure:

$$1. \sigma_0 = \underbrace{\bar{\sigma}_{0,S}}_b + \underbrace{(\bar{\sigma}_{0,L} - \bar{\sigma}_{0,S})}_a f + \epsilon \quad [3]$$

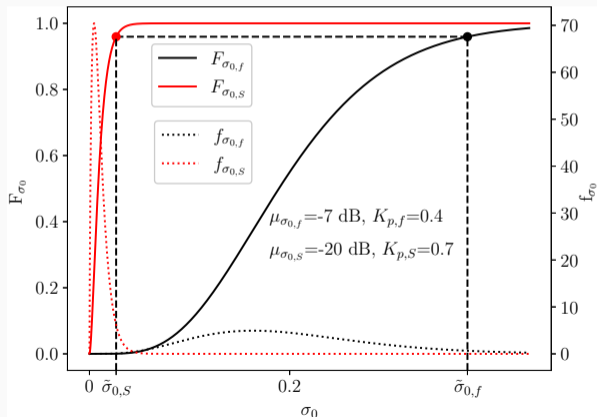
$$2. \bar{\sigma}_{0,f} = af + \bar{\sigma}_{0,S}$$

$$3. \bar{\sigma}_{0,f} \rightarrow \hat{K}_p(\bar{\sigma}_{0,f})$$

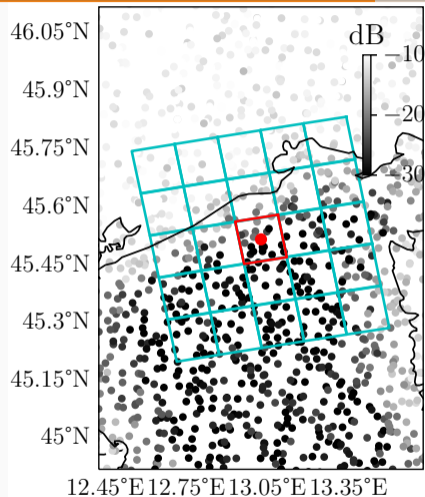
4. CDF matching:

$$F_{\bar{\sigma}_{0,f}, \hat{K}_p}(\tilde{\sigma}_{0,f}) \equiv F_{\bar{\sigma}_{0,S}, \hat{K}_p}(\tilde{\sigma}_{0,S})$$

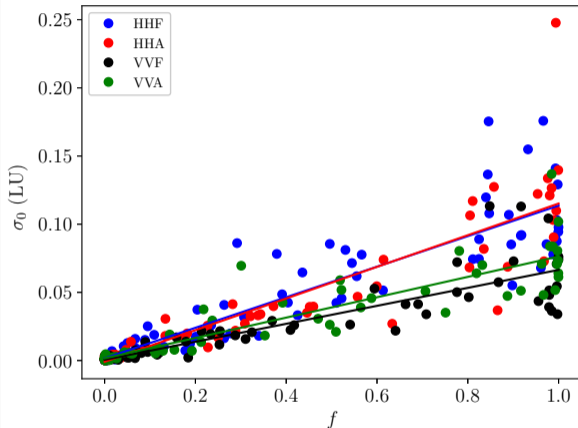
[3] J. Vogelzang, A. Stoffelen, "ASCAT land correction",  
SAF/OSI/CDOP3/KNMI/TEC/TN/384



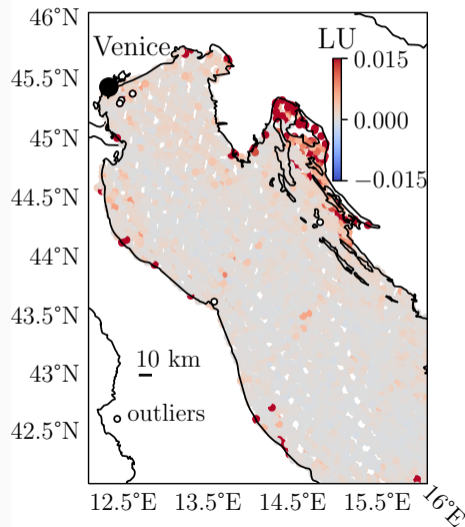
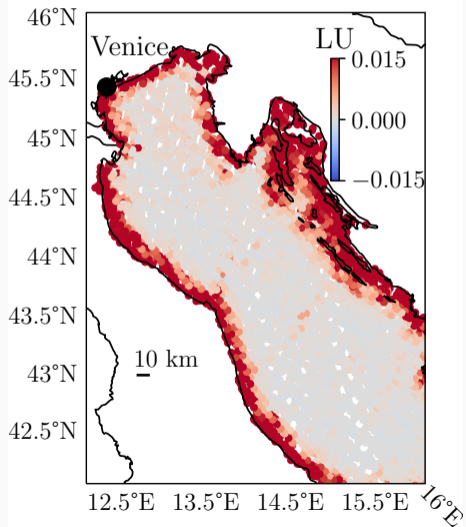
# Area test: north Adriatic (Mediterranean basin)



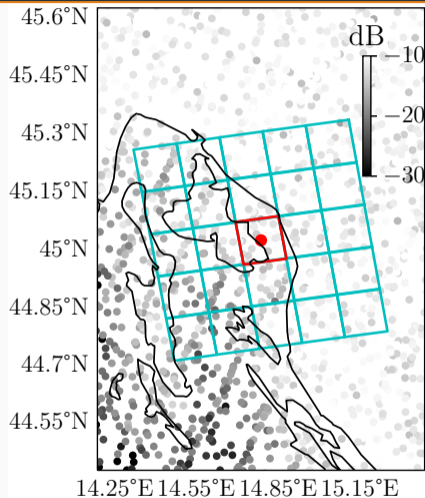
QuikSCAT Full-Resolution file ID 40653



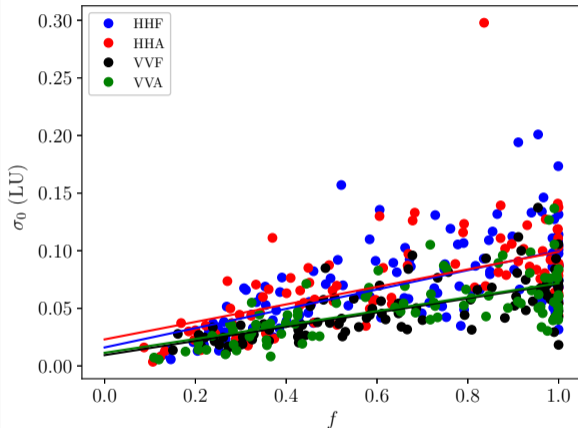
# corrected $\sigma_0$ with noise regularization



# Residual land contamination due to lack of $\sigma_{0,S}$



QuikSCAT Full-Resolution file ID 40653



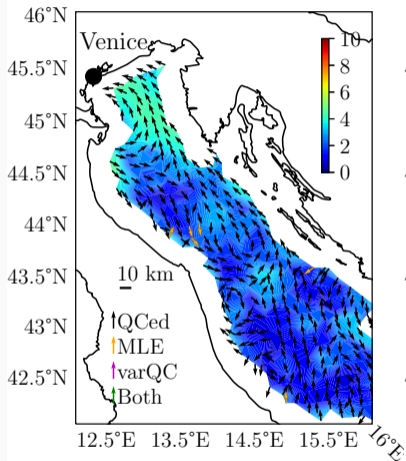
# Retrieval experiments

Name	$f^{th}$	Noise regularization	Orbit IDs
CTRL	0.02	NO	40651-40664
NC	0.5	NO	40651-40664
NR	0.5	YES	40651-40664

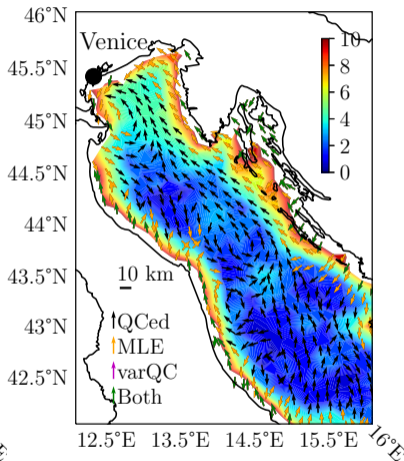
CTRL represents the state-of-the-art at OSI-SAF

- Two area tests: **North Adriatic** (Mediterranean) and **Netherlands**
- Day: 10<sup>th</sup> April 2007
- WVC grid: 12.5 km

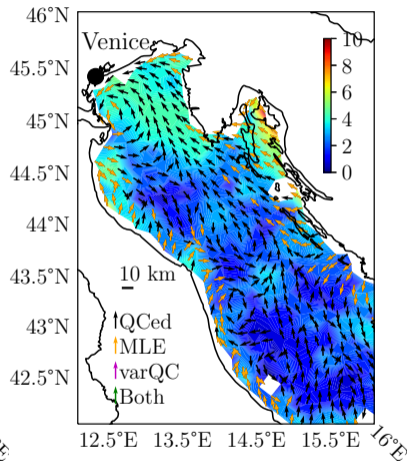
# Retrieved winds: North Adriatic



**CTRL**

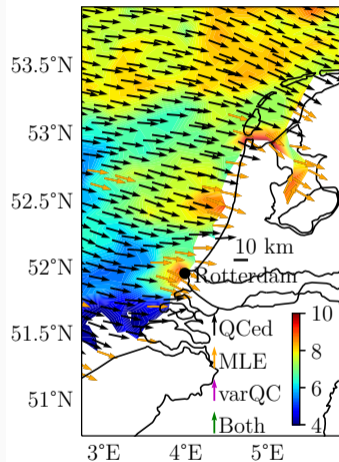
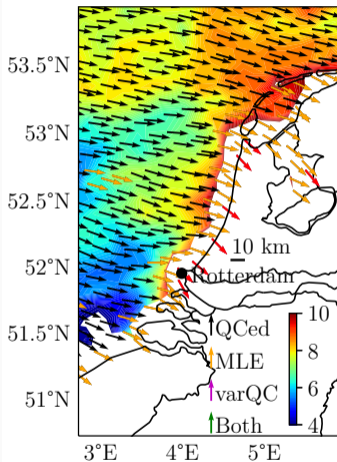
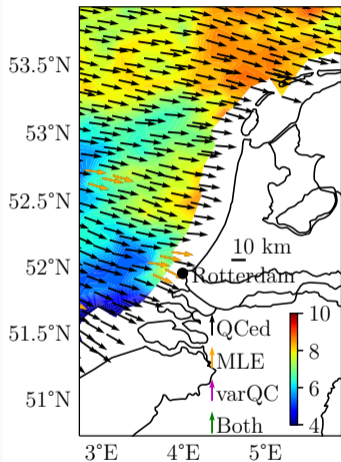


**NC**



**NR**

# Retrieved winds: Netherlands



**CTRL**

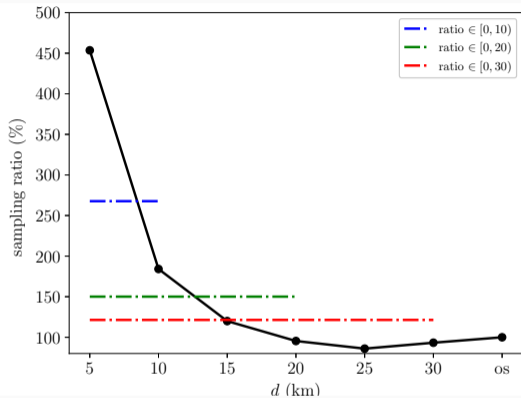
IOVWST 2023, Nanjing, 30<sup>th</sup> Nov

**NC**

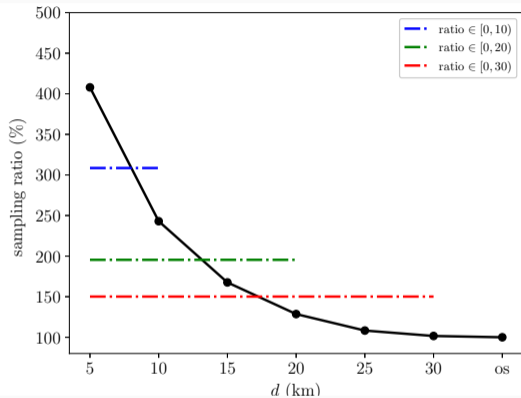
**NR**



# Sampling rate improvement



w.r.t. CTRL

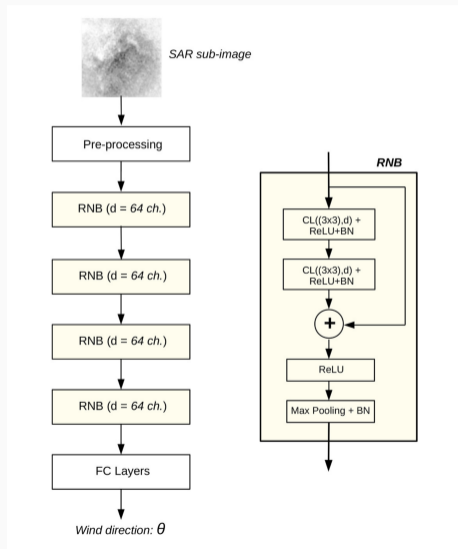


w.r.t. NC

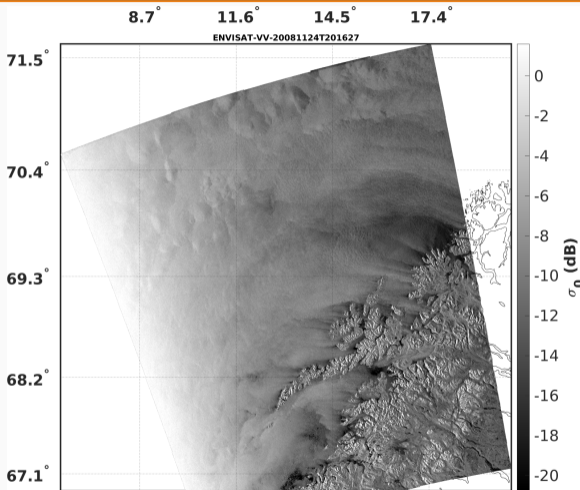
# Envisat ASAR-derived winds

- Wind direction:
  - Method: Deep Learning with Convolutional Neural Network (CCN) with Residual Neural Network structure (**ResNet**)
  - Configuration: 4 layers, 64 channels
  - Training dataset of **816,000** pairs:
    - 25 Sentinel 1 GRDW IW images
    - ECMWF FC 0.125°
- Wind speed:
  - $U=C\_SarMod2^{-1}(\sigma_0, \theta)$  (Lu et al. 2018)

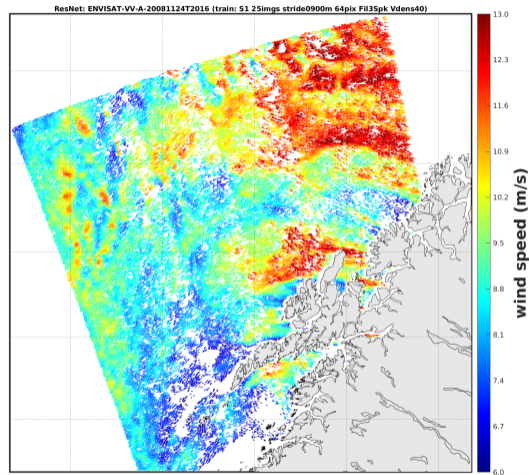
Zanchetta and Zecchetto, "Wind direction retrieval from Sentinel-1 SAR images using ResNet", Remote Sensing of Environment, 253, 2021 (<https://doi.org/10.1016/j.rse.2020.112178>)



# ASAR-derived winds

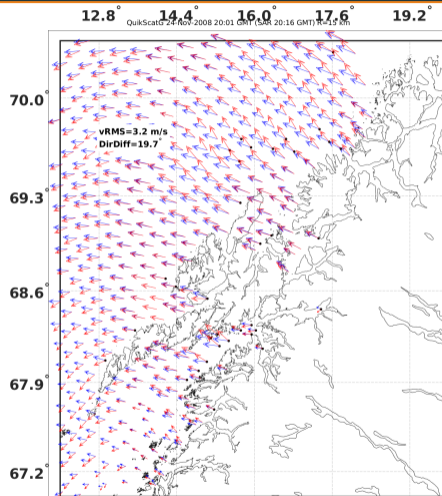
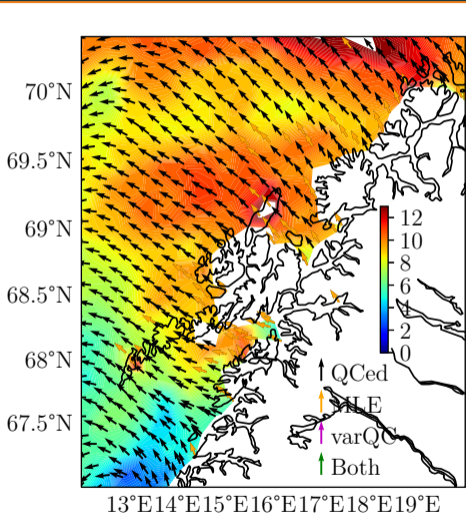


$\sigma_{0,ASAR}$



$\vec{u}^{0.9}_{ASAR}$

# Comparison between QuikScat and Envisat derived winds



$\vec{u}_{ASAR}^{12.5}$  vs  $\vec{u}_{QS}^{12.5}$   
Black markers:  
rainy WVCs

# Preliminary Conclusions & Future Work

## Preliminary conclusions

- $\sigma_0$  correction with noise regularization is effective
- Wind retrievals are good, also in internal seas
- Coastal sampling gain: +400% within 5 km and  $\approx 300\%$  within 10 km
- Encouraging agreement with SAR-derived winds

## Future work

- MLE threshold tuning
- Assessment of any residual contamination
- Validate winds (how? Against buoys, models, SAR-derived winds?)
- Improve ResNet-derived winds and consistency with QS winds
- Export Noise-Reg to other pencil-beam scats (OceanSat, HY-2)

# Acknowledgements

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OSI\_VSA\_20\_03, OSI\_VSA\_21\_03,  
OSI\_VSA\_22\_02
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Dumbar (JPL)



# Back-up slides