

# Evaluation and Applications of Ocean Surface Wind Product From Fengyun-3 and Tianmu-1 GNSS-R Constellations

Feixiong Huang<sup>1</sup>, Cong Yin<sup>1</sup>, Xiaochun Zhai<sup>2</sup>, Rongrong Li<sup>3</sup>, Lei Li<sup>3</sup>

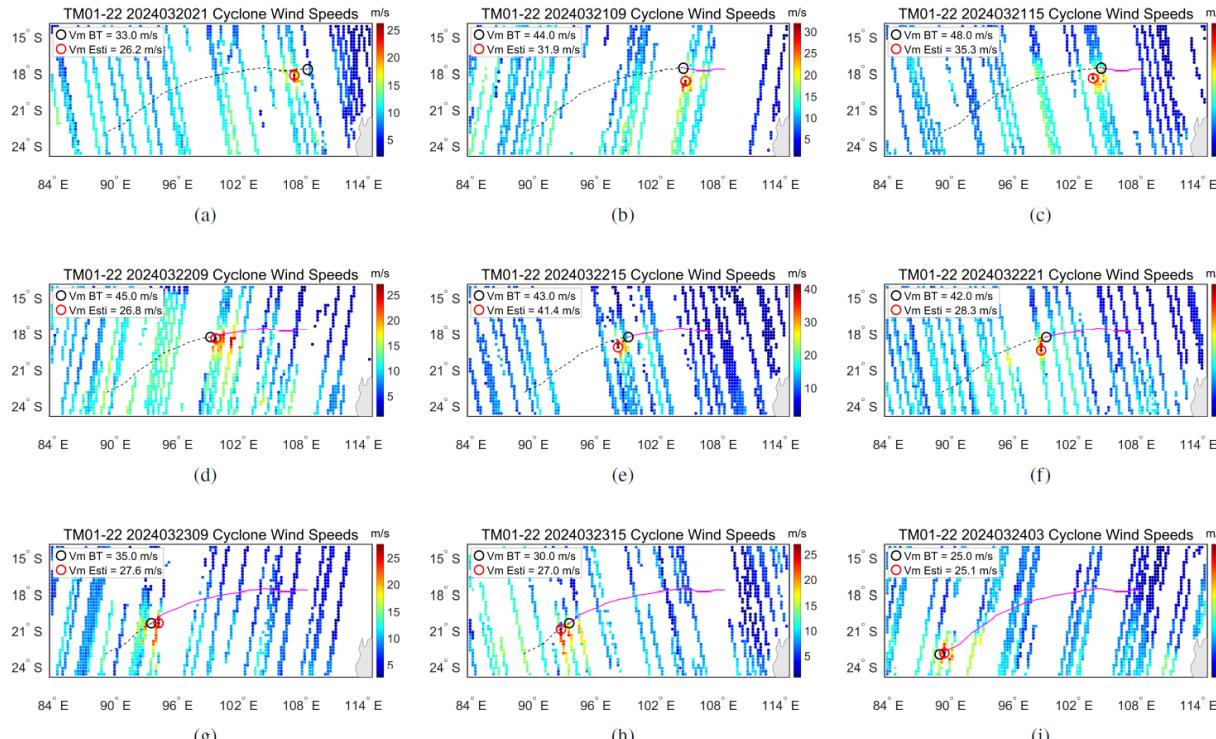
<sup>1</sup>National Space Science Center, Chinese Academy of Sciences  
<sup>2</sup>National Satellite Meteorological Center, China Meteorological Administration  
<sup>3</sup>Aerospace Tianmu (Chongqing) Satellite Science and Technology Co., Ltd.

GNSS-R Ocean Surface Wind Data (25 satellites):

Mission	# Satellites	Launch Year
FY-3E/F/G	3	2021-2023
Tianmu-1	22	2023-2024

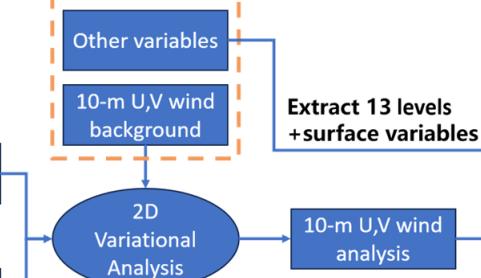
high temporal resolution

**Application 1:** Monitoring the **location** and **intensity** of TC Neville in March 2024 by **every 6 hour**



**Application 2: Improving TC track forecasting of AI weather models**

NCEP analysis (37 levels)



Pangu AI Weather Model (13 levels, 13\*5+4=69 variables)

Type	Full name	Abbreviation
geopotential	Z	
specific humidity	Q	
temperature	T	
<i>u</i> -component of wind speed	U	
<i>v</i> -component of wind speed	V	
mean sea level pressure	MSLP	
2m temperature	T2M	
<i>u</i> -component of 10m wind speed	U10	
<i>v</i> -component of 10m wind speed	V10	

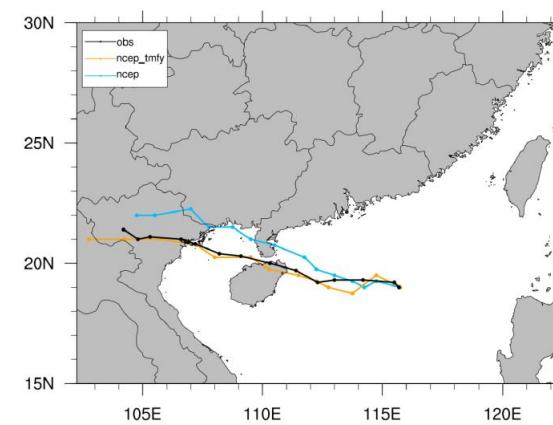
Forecast

usaobs: TC best track (IBTrACS)

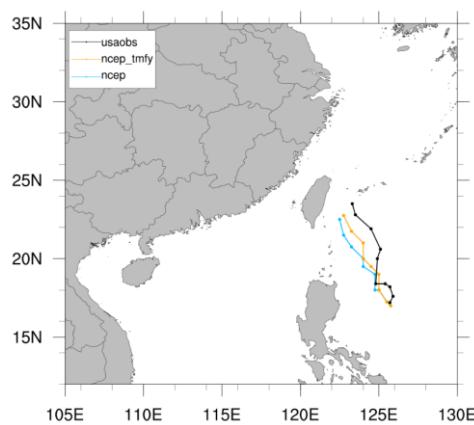
ncep: use NCEP analysis as input

ncep\_tmfy: NCEP analysis and GNSS-R winds as input

Typhoon Yagi  
(20240905 00Z)



Typhoon Gaemi  
(20240721 18Z)



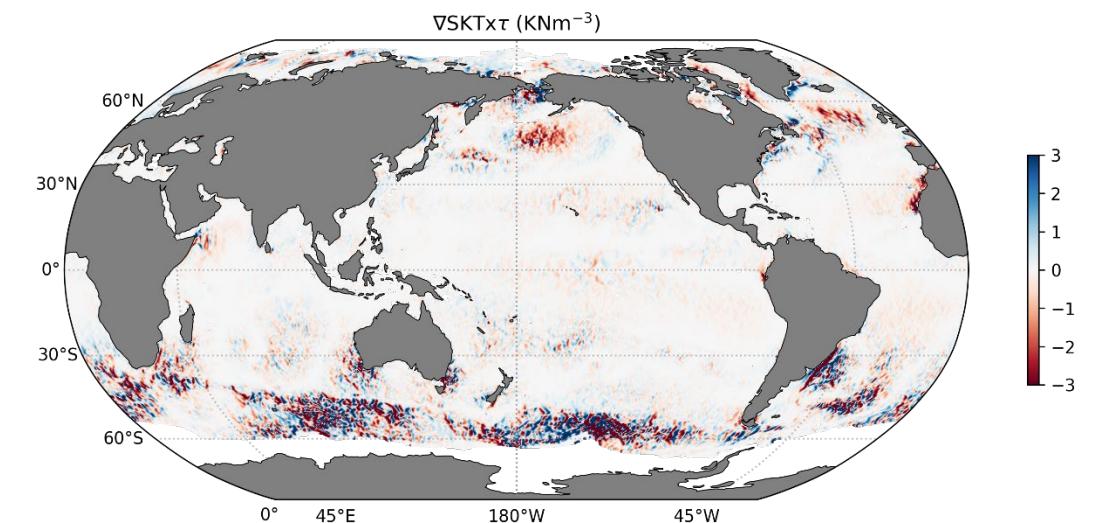
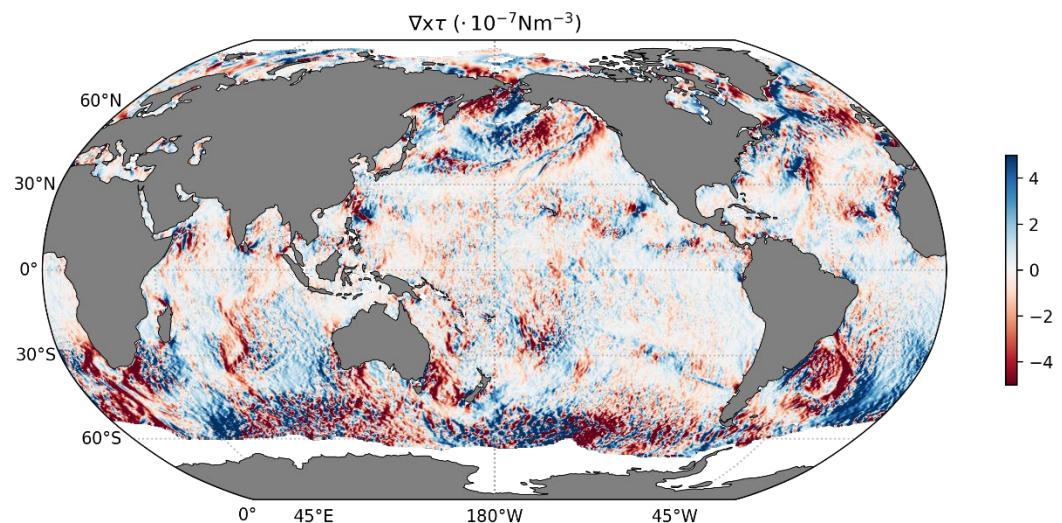


# Coupling Atmospheric Dynamics and Ocean with Winds from Satellites

*Stavroula Biri, Ad Stoffelen*

How can we further reduce systematic model errors on large scale and atmospheric mesoscale?

- › Evaluate the missing variance and error variance in NWP model wind fields; attribute these to geophysical processes using physics-informed diagnostics



# Stokes drift and current coupling at the submesoscale

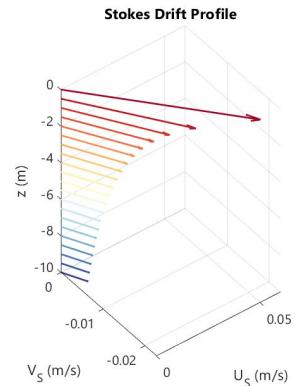
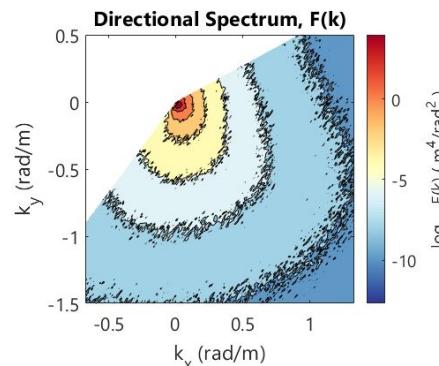
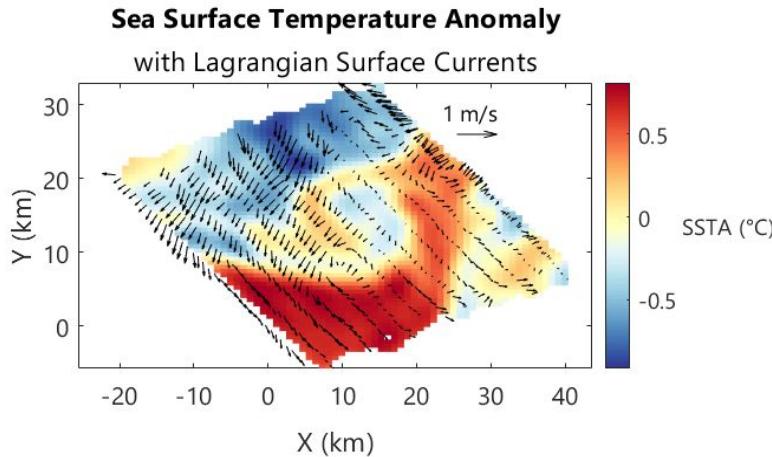
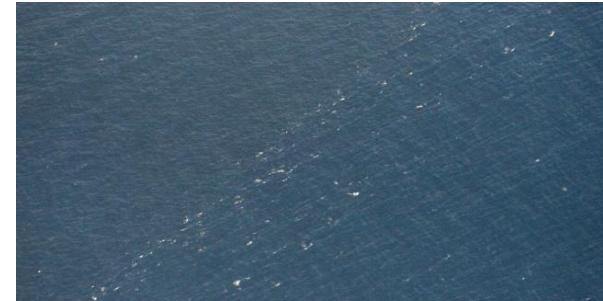
Kayli Matsuyoshi<sup>1</sup>, Luc Lenain<sup>1</sup>, Mara Freilich<sup>2</sup>, Nick Pizzo<sup>3</sup>

<sup>1</sup>Scripps Institution of Oceanography, University of California, San Diego    <sup>2</sup>Brown University

<sup>3</sup>University of Rhode Island, Graduate School of Oceanography



Determining the **role of waves in submesoscale feature evolution** by investigating whether Stokes forces induce, respond to, or are independent of the velocity gradient fields generated.



# MODELING SPRAY IMPACTS ON WIND SPEED, STRESS, AND DRAG COEFFICIENTS

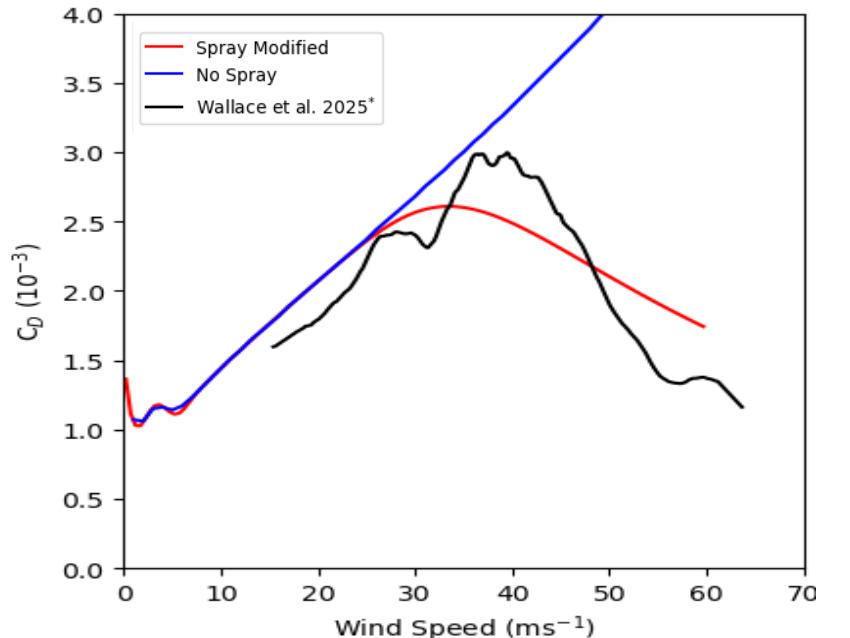
Renee J. Richardson<sup>1,2</sup> and Mark A. Bourassa<sup>1,2</sup>

Florida State University<sup>1</sup>, Department of Earth, Ocean and Atmospheric Science, Tallahassee, FL, USA  
Center for Ocean-Atmospheric Prediction Studies<sup>2</sup>, Tallahassee FL, USA

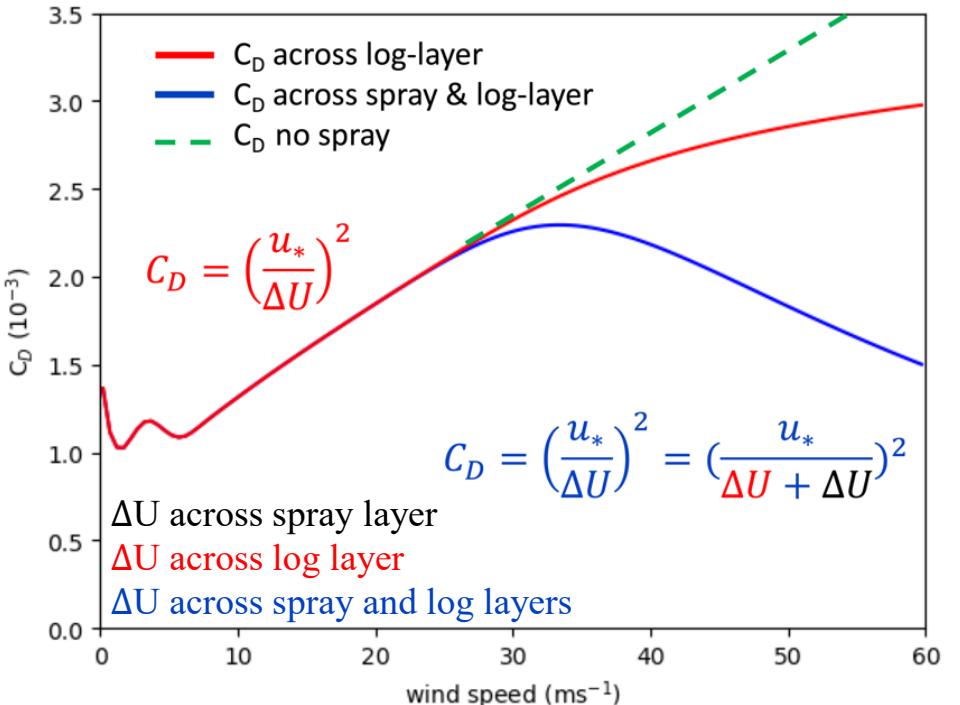


IOVWST Meeting 2025  
Darmstadt, Germany

- Nontraditional boundary layer approach assumes a thin, spray layer being present underneath the constant flux layer
- Under this assumption,  $C_D$  is more consistent with observational analyses, however, more processes are at play

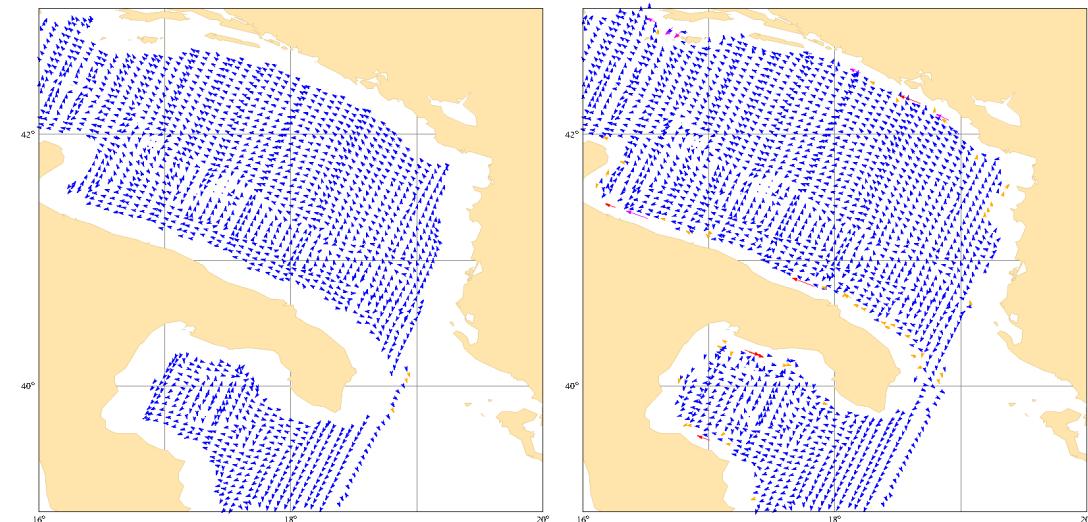


- Total vertical wind shear across the surface layer ( $\Delta U$ ) demonstrate a  $C_D$  that is more representative of observations

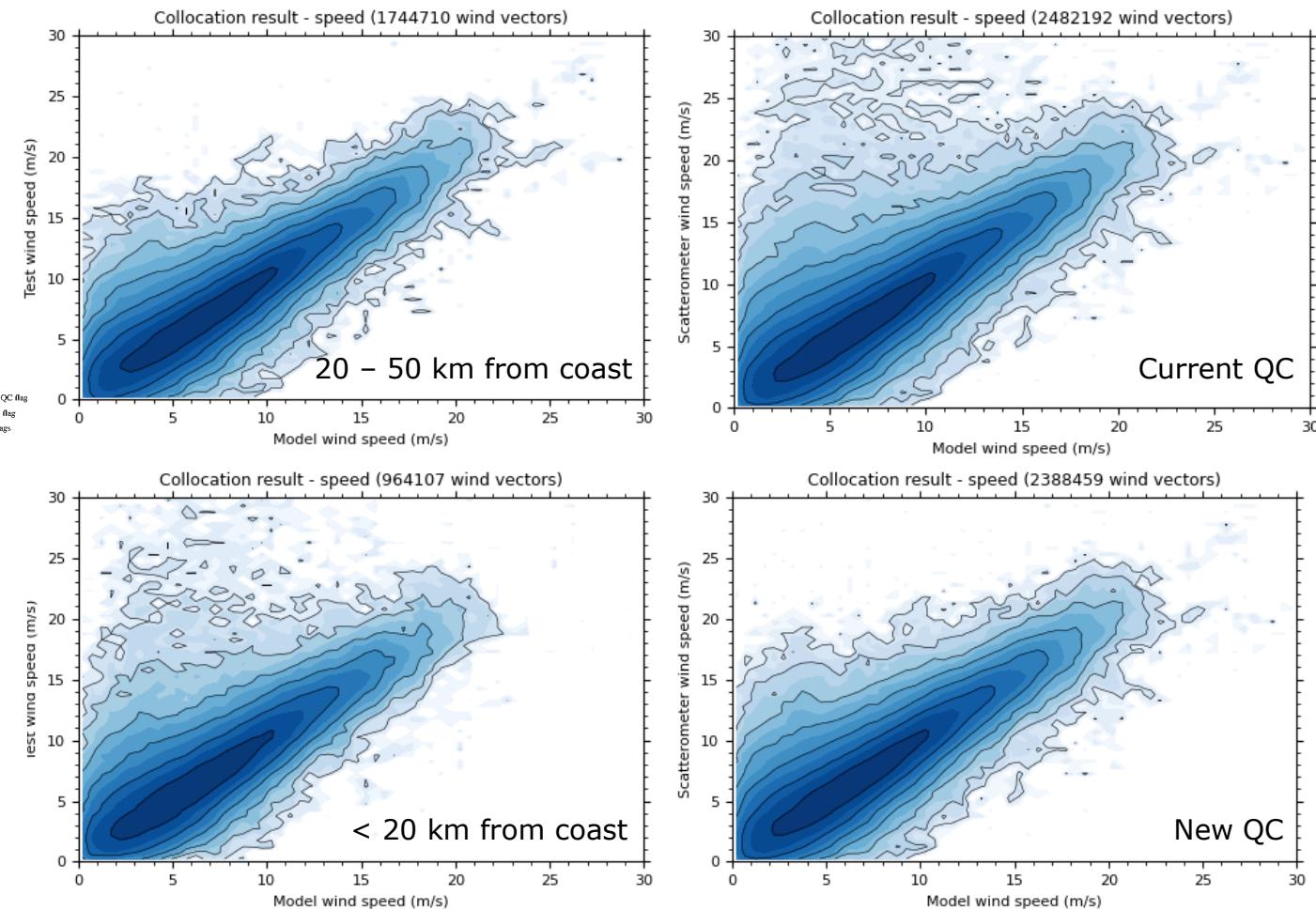


# ASCAT higher resolution winds closer to the coast

- › Left column: we have implemented land corrections and coastal retrievals in the 6 km ASCAT wind product
- › Mid column: this comes with some unrealistic high wind speeds close to the coast
- › Right column: we present some ways forward to improve the situation



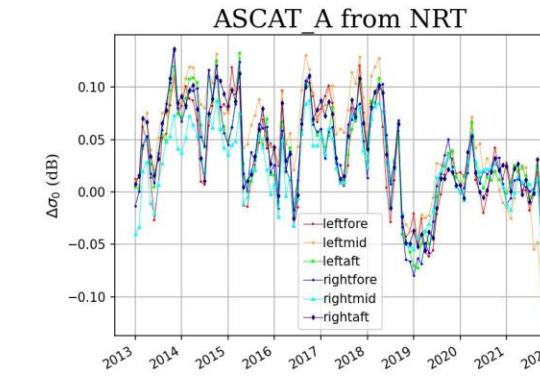
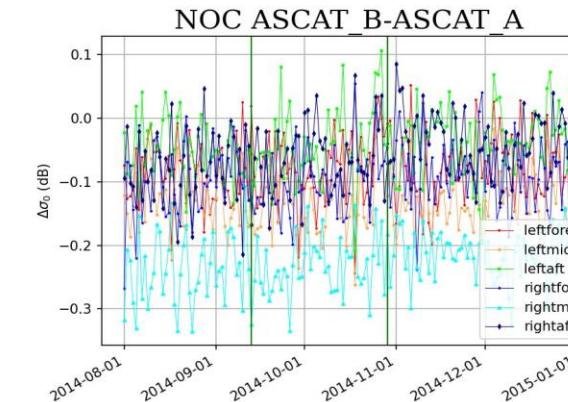
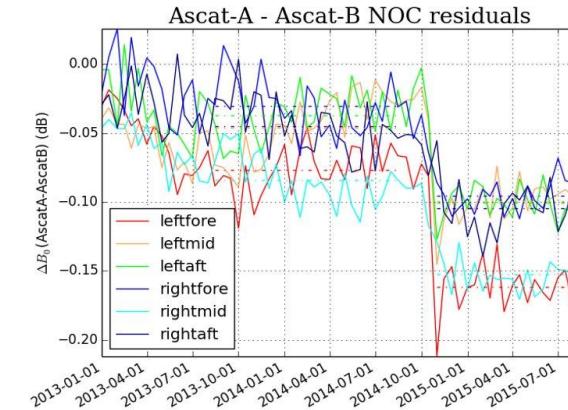
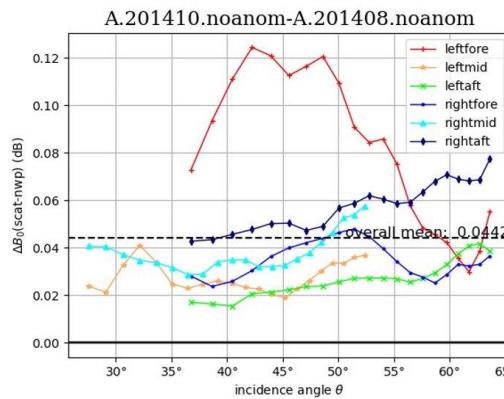
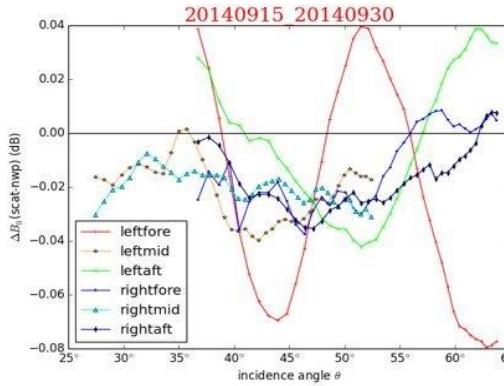
A. Verhoef, J. Verspeek, A. Stoffelen, KNMI – IOVWST 2025



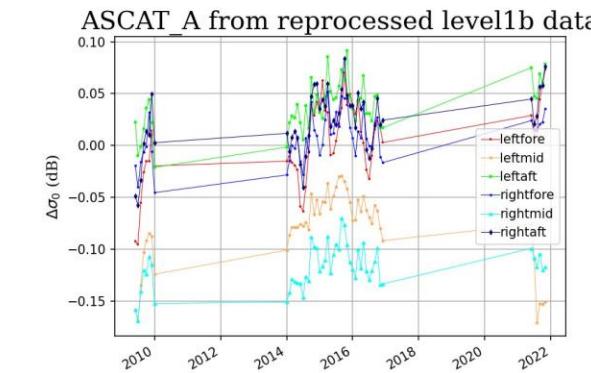


# ASCAT-A scatterometer wind product reprocessing

- ASCAT-A scatterometer reprocessing planned with ERA5 stress-equivalent 10 m background winds
- The level1b data has been reprocessed by EUMETSAT
- Instrument anomalies have been re-evaluated



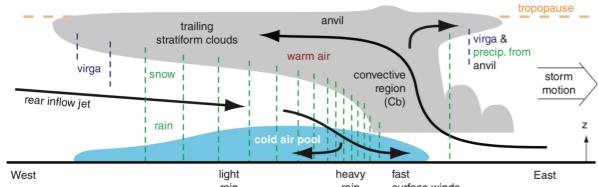
NRT



Reprocessed  
level1b test data

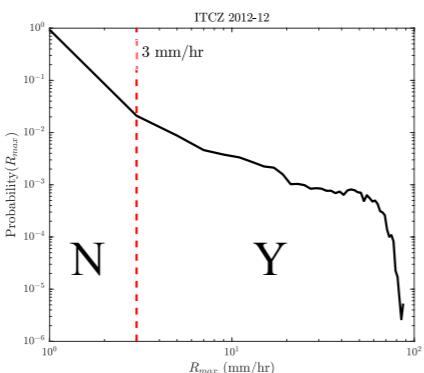
# Identifying Tropical Cold Pools with Singularity Exponents

Gregory P King<sup>(ATTIC)</sup>, Marcos Portabella<sup>(ICM)</sup> and Wenming Lin<sup>(NUIST)</sup>



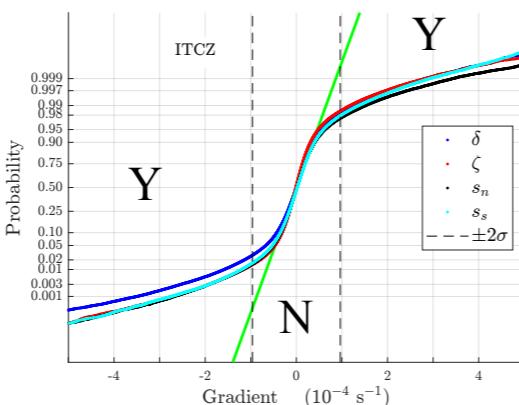
**Rain rate extremes**

$$R_{max} > 3 \text{ m/hr}$$



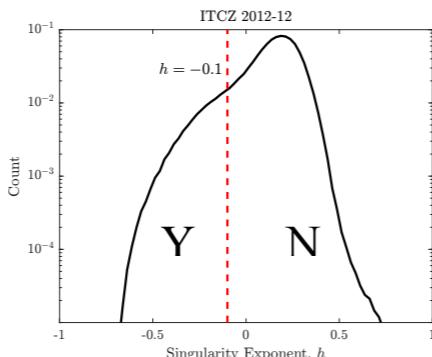
**Wind gradients**

$$|\nabla \mathbf{u}|_F = \frac{1}{2} \sqrt{\delta^2 + \zeta^2 + s_n^2 + s_s^2}$$



**Singularity Exponents**

$$|\nabla \mathbf{u}(\mathbf{x})| \sim r^{h(\mathbf{x})}$$



Contingency table for binary variables  $A$  and  $B$ .  
 $Y \Rightarrow$  exceeds threshold.

		$B$	
		$Y$	$N$
$A$	$Y$	$n_{11}$	$n_{12}$
	$N$	$n_{21}$	$n_{22}$

$$\text{Odds Ratio} = \frac{n_{11} n_{22}}{n_{12} n_{21}}$$

**Significant correlation when**  
 $OR(A, B) > 1$

**Case showing spatial locations of**

$$\{n_{11}, n_{12}, n_{21}, n_{22}\}$$

