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## Wind Bias Correction Guide

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## Scatterometer missions overview (WMO OSCAR)

Copernicus Vianine Service

Instrument	NRT?	Relevance	Satellite	Orbit	DLR	2022	2023	2024	2025	2026	2027	2028	2029	2030	2031	2032	2033	2034	2035
WindRAD		1 - primary	FY-3E	05:40 desc		X	х	X	X	X	X								
WindRAD		1 - primary	FY-3J	05:00 desc							х	X	X	x	X	X	X	X	X
ASCAT	Yes	2 - very high	Metop-B	09:31 desc	50	Х	Х	Х											
ASCAT	Yes	2 - very high	Metop-C	09:31 desc	85	х	х	х	х	х	х								
SCA (Scatterometer)		2 - very high	Metop-SG-B1	09:30 desc					х	x	x	х	х	х	х	х			
SCA (Scatterometer)		2 - very high	Metop-SG-B2	09:30 desc												x	х	X	x
CSCAT 🕕		2 - very high	CFOSAT	07:00 desc		Х	x	1											
HSCAT		2 - very high	HY-2B	06:00 desc	273	х	х												
HSCAT		2 - very high	HY-2D	66 °		х	х	х	х	х									
HSCAT		2 - very high	HY-2E	06:00 desc				х	x	х	х	х							
HSCAT		2 - very high	HY-2C	66 °		х	х	х	х										
HSCAT		2 - very high	HY-2F	66 °					x	х	х	х	х						
OSCAT-3		2 - very high	OceanSat-3 (EOS-0	6 12:00 desc			x	x	х	х	х	х	х	х					
Source: https://spa	ce.oscai	r.wmo.int/gapar	nalyses?mission=1	2															

#### Past C-band missions :

ERS-1,2/ESCAT MetOp-A/ASCAT 10:30 desc. 9:30 desc.

1992-1996, 1995-2000 2007-2021

> Prepare yourselves for many scatterometers ③

#### Past Ku-band missions :

 SeaWinds/QuikScat
 6:00 desc.
 1999-2009

 RapidScat/ISS
 52 \*
 2014-2016

 OceanSat-2/OSCAT-1
 0:00 desc.
 2009-2014

 ScatSat-1/OSCAT-2
 8:45 desc.
 2016-2021

# **Quintuple collocation analysis**

<b>Observing System</b>	$\sigma_u$ (m/s)	$std(\sigma_u)$ (m/s)	$\sigma_v$ (m/s)	$std(\sigma_v)$ (m/s)			
buoys	0.914	0.017	1.063	0.020			
ASCAT-A (C-band)	0.372	0.022	0.505	0.029			
ASCAT-B (C-band)	0.390	0.025	0.444	0.020			
ScatSat (Ku-band)	0.683	0.018	0.594	0.021			
ECMWF	0.845	0.017	1.006	0.021			

Table 2. Observation error standard deviations and their accuracies.

- Beyond triple and quadruple collocation for global calibration and random error assessment
- Consolidated several methodologies to solve collocation error equations
- Added better ability to approximate the errors of the errors
- Confirms the excellent accuracy of scatterometer winds
- Stress-equivalent 10-m winds

#### Jur Vogelzang and Ad Stoffelen, 2022a, 2022b, 2021

## Rationale

- ✓ See SCA SAG science plan (drafted 2016)
- ✓ U10 Model <u>biases</u> are locally rather high compared to innovation, violating Best Linear Unbiased Estimate paradigm in data assimilation
- ✓ A few decades of model improvement have not solved this problem, though one is still trying actively; it is a problem for ocean forcing too
- ✓ The <u>EU Copernicus Marine Service L4 OPS</u> and <u>ERA5</u> corrections can be inversely applied to the scatterometer winds to adjust them to be geographically unbiased with respect to the model
- ✓ ECMWF provided a reference run without scatterometers for which NUIST and KNMI computed model biases, averaging over 20 days (like for Copernicus L4 product)
- ✓ NUIST applied these biases to obtain adjusted SCAT BUFR products
- ✓ ECMWF will run a SCAT\* OSE and compare it to reference OSEs with (SCAT OSE) and without (noSCAT OSE) scatterometer data assimilation
- ✓ EUMETSAT MIDAS project result on scatterometer OSEs with the HARMONIE model also points to a bias problem



### L3 daily Copernicus Marine Service



L4 hourly Copernicus Marine Service



## **Quality Control and ocean winds**

- EUMETSAT OSI SAF continuously improves quality control (QC)
- For Ku-band scatterometers we need to control rain events (=>)
- Develop algorithms to correct for remaining observational sampling biases
- Downbursts in moist convection have a large and systematic impact on air-sea interaction
- These fast (30 min) and mesoscale (few km) processes are not well tracked in global NWP
- Scatterometers help correct climatological biases due to missing processes in models

#### Zhao et al., 2023

King et al., 2022

Xu and Stoffelen, 2021, 2019

Trindade et al., 2020, 2023

Belmonte and Stoffelen, 2019







## SCAT\* : Making ASCAT and HY2 consistent

- ✓ Significant inconsistencies of wind speed are found between ASCAT and HSCAT NRT products; Zhixiong Wang confirmed that this is NOT caused by resolution differences (25/50km)
- ✓ Rep01: By using more accurate σ<sup>0</sup> NWP Ocean Calibration (NOC) after improved rain QC, winds among HSCAT-B, C, and D become more consistent, but NOT close enough to ASCAT
- ✓ Rep02: By making and using the new NSCAT-4ds.hy2 GMF and compute corresponding NOC, winds from HSCAT and ASCAT show good agreements. However, wind speeds below 2 m/s or above 20 m/s still need further calibration after running over longer periods
- ✓ The products of ASCAT NRT and HSCAT Rep02 are the best version choices as sea surface wind vector inputs to the ECMWF ASCAT\* OSE
- The residual biases (i.e., depending on instrument or WVCs) are acceptable, and we can move on to the next step

More details are given on following slides . . .

## Collocated ASCAT and HSCAT winds

### **HSCAT NRT**

8 10 12 14 16 18 20 22 24 26 28 30

Average wind speed (m/s)

HSCATC - ASCATB

HSCATD - ASCATB

- HSCATC - ASCATC

HSCATD - ASCATC

1.0

0.8

0.6

0.4

0.2

0.0

-0.2

-0.4

-0.6

-0.8

-1.0

0 2

4 6

wind speed bias (m/s)



- NWP Ocean Calibration (NOC)

### HSCAT Rep02



- Improve rain QC
- NOC
- Improve Geophysical Model Functions 7

### ◆ Spatial distance ≤ 50\*0.7071km

• Time diff.  $\leq$  45min

## Verifying with ECMWF winds

☑ It is clear that: HSCAT Rep02 is better.

☑ Wind speed dependent wind seed biases are reduced, and the curves of HSCAT become more similar to ASCAT curves.







#### **HSCAT-BNRT**

#### HSCAT-B Rep01

### HSCAT-B Rep02



### Wind speed biases of SCA - NWP



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### Wind direction biases of SCA - NWP



HSCAT-C Rep02

HSCAT-B Rep02

HSCAT-D Rep02

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## Meridional (v) model bias adjustment



## Zonal (u) model bias adjustment



- Top: large u first guess biases, both in runs with (OPS) / without (OSE1) ASCATs and HY2B
- Bottom: ASCATB\_SC is adjusted to OSE1 and not to ECMWF\_OPS, while with small biases
- Top: large u first guess biases, both in model runs with/without ASCATs and HYB
- Bottom: ASCATB\_SC is well adjusted to OSE1
- Top: large u biases in ASCAT-B\_SC as expected
- Bottom: ECMWF\_OPS minus OSE1 is complement of ASCAT-B\_SC minus OPS (on left)
- OPS FG biases adjust only a little to the scatterometers



Mesoscale Improved Data Assimilation of Scatterometer winds (MIDAS)

## MIDAS conclusions

- HARMONIE 3-hour 4D-Var better than the widely used 3D-Var
- ASCAT improves the forecast skill both in 3D- and 4D-Var
- Tested data thinning distances, superobbing and observation error inflation
- Particular effects on the v component
- Error inflation at full density similar to superobbing statistically (as expected in <u>Stoffelen et al., 2020</u>)
- Local model biases are substantial with respect to the innovations and violate the data assimilation BLUE paradigm
- Scatterometer winds are not effective to initialise dynamical weather features and model biases need to be accounted for to better exploit scatterometer winds in HARMONIE

This was a EUMETSAT study



## Mesoscale Improved Data Assimilation of Scatterometer winds (MIDAS)



Median OSCAT – noSCAT FG 000 midas4DVar ctrl OSCAT.csv u10 median bias on a 0.5 X 0.5 d Period: 07 Feb 2020 - 15 Mar 2020 u10 5 4Dvar 5ºW

2.2

1.4

0.6

0.0

-0.6

-1.4

-77



Draft version prepared by: Isabel Monteiro, Gert-Jan Marseille, Fabíola Silva, Jan Barkmeijer and Ad Stoffelen

## Conclusions

- Model biases of 10-m stress-equivalent wind (U10s) are substantial with respect to observations
- Scatterometers can map out the rather stable spatial biases well
- Biases prevent effective data assimilation (BLUE paradigm)
- Experiment with ECMWF o-b bias correction in progress by adjusting scatterometer BUFR data
- Biases also prevent effective scatterometer data assimilation in HARMONIE
- U10s biases affect ocean forcing and hence air-sea coupling and earth system dynamics (ocean is 70% of the surface)
- EUMETSAT awarded a fellow position at KNMI/ICM/ECMWF to address data assimilation, ocean forcing and physical causation of biases
- EUMETSAT OSI SAF visiting scientist Evgenia Makarova at ICM employs Machine Learning based on model parameters to predict the biases (MOS)
- Each scatterometer may contribute a few % in the reduction of the forecast errors and with 6 complementary scatterometers it may be a worthwhile investment to improve their assimilation by addressing remaining problems, of which model biases is a prominent one
- Furthermore, scatterometers can be well exploited to (much?) improve the coupled model dynamics at the air-sea interface



### Brief Introduction of Datasets

✓ ASCAT-B 25km NRT✓ ASCAT-C 25km NRT

✓ HSCAT-B 50km NRT (NOC: +0.62(HH), -0.63(VV))
 ✓ HSCAT-C 50km NRT (NOC: -1.17(HH), -1.32(VV))
 ✓ HSCAT-D 50km NRT (NOC: -0.34(HH), -0.12(VV))



♦ NWP data are taken from BUFR files, i.e., the same as NRT processing used!

- ◆ Time period: Dec. 01, 2021 ~ April 30, 2022
- ◆ SST data are taken from ERA5 at analysis time.

♦ NSCAT-4ds.hy2 GMF was made using CDF matching tech. based on collocated ascatb and hscatc+d winds

• New NOC was calculated using NSCAT-4ds.hy2 GMF and NWP winds contained in BUFR files.

### HSCAT-B NRT

### HSCAT-B Rep01

### HSCAT-B Rep02

c

6

- 3

- 0

-3

-6



### HSCAT-C NRT

### HSCAT-C Rep01

### HSCAT-C Rep02



### HSCAT-D NRT

### HSCAT-D Rep01

### HSCAT-D Rep02







#### **HSCAT-C NRT**

#### HSCAT-C Rep01

### HSCAT-C Rep02



#### **HSCAT-D NRT**

#### HSCAT-D Rep01

#### HSCAT-D Rep02

