

Royal Netherlands Meteorological Institute Ministry of Infrastructure and the Environment

Mesoscale scatterometer assimilation

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Do we have enough data?



Does Dynamical Downscaling With Regional Climate Models add Value to Surface Marine Wind Speed From Reanalyses?

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Simulations with RCMs REMO and CLM: (available from CRASTDat Database)

•Three hindcasts with RCMs REMO (Jakob and Podzun, 1997) and CLM (Böhm et al. 2006) •Initialization and forcing at lateral boundaries: NCEP/NCAR-Reanalysis (NRA), ~1.875° resolution, •SN-REMO & CLM hindcasts are additionally forced by spectral nudging (von Storch et al., 2000)

Hindcast	STD-REMO (Standard)	SN-REMO	CLM
Based on:	EM	EM	LM
	Hydrostatic	Hydrostatic	Non-hydrostatic
Forcing:	NRA	NRA	NRA
Spectral Nudging:	No	Yes	Yes
Resolution:	0.5°	0.5°	0.44°

 For that purpose a gridded QuikSCAT Level 2B 12.5 km swath (L2B12) data set is produced on SN-REMO grid (rain flagged L2B12 data discarded) co-location with SN-REMO: QuikSCAT wind speed retrieval max. 12.5 km and +/- 10 min from SN-REMO grid point / time step

• Modified BSS = $\begin{cases} 1 - \sigma_F^2 \sigma_R^{-2} & \text{if } \sigma_F^2 \le \sigma_R^2 \\ \sigma_R^2 \sigma_F^{-2} - 1 & \text{if } \sigma_F^2 > \sigma_R^2 \end{cases}$

• "Forecast" F: SNREMO, reference "forecast" R: NRA, predictand/observation: gridded QuikSCAT L2B12 data



HARMONIE from ECMWF

- HSCAT scatterometer 50 km
- HARMONIE effective resolution 25 km, grid 2.5 km

	(m/s)	bias u_{10m}	stdev u_{10m}	bias v _{10m}	stdev v _{10m}
	HSCAT	(23.961	collocations); $\overline{\Delta}$	$\overline{t} = -0.29; \ \overline{ 2 }$	$\Delta t = 0.85$
	(o-b)	-0.46	1.61	-0.24	1.57
Temporal interpolation:	$(o - b_t)$	-0.46	1.36	-0.22	1.29
+ spatial averaging:	$(o - \bar{b}_t)$	-0.45	1.25	-0.22	1.18

ECMWF:		t_f	bias u_{10m}	stdev u_{10m}	bias v _{10m}	stdev v _{10m}
	HSCAT	5.6	-0.11	1.09	0.05	1.15

- ECMWF 6-hour forecast better than matched 50-km scale time-interpolated HARMONIE background
- \succ ECMWF resolution is ~150 km over the open ocean
- Deterministic resolution

Nastrom & Gage Observed Spectrum



- Tropospheric spectra are close to k^{-5/3}
- 3D turbulence
- L/H ~ 100
- SD = 0.4 (log spectral density)
- Least variance/ detectability in small scales



Nastrom & Gage Observed Spectrum



Small-scale data assimilation

- The amplitude spectrum of small-scale atmospheric waves can be well simulated in NWP models, but the determination of the phases of these waves will be problematic in absence of well-determined forcing (orography) or observations
- Undetermined phases at high resolution cause
 - Increased NWP model error, B' > B
 - Model errors get more variable and uncertain since small scales tend to be coherent; coherence is of most interest
 - B error structures will be spatially more sharp
 - Increased O-B, while the observation (representativeness) errors will be reduced; observations (should) get more weight, O' < O
 - Increments would be larger
 - When O' > B, the analysis error will be larger too ! A' > A

Challenges

- > Adaptive B covariances are difficult to estimate
- More (wind) observations are needed to spatially sample small-scale B structures
- \succ Observations need to be accurate, O < B
- How to prevent overfitting (uncertain B, small O) due to inaccurate and high innovation weights ?
- And spin-up due to more noisy analysis (statistical B) ?
- Separate determined from undetermined scales in data assimilation, e.g.,
 - > Data assimilation with ensemble mean ?
 - Maintain broad B ?
 - SuperMod up to determined scales ?



Examples

Estimated B error variances





ECMWF Ensemble Data Assimilation (EDA background error) ASCAT-derived ECMWF background error by triple collocation in QC classes

NWP Background spatial error correlation structure



Cyclone SH mm/h



Number of ambiguities

Cyclone SH, 2DVAR analyses



Cyclone SH, selected solutions



All the QC-accepted data (March-August 2009)

	ASCAT-ECMWF-buoy comparison (mean buoy winds)					
	ASCAT vs ECMWF	ASCAT vs buoy point wind	Ν			
Default	2.27	1.86	6908			
New	2.26	1.83				

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	ASCAT-ECMWF-buoy comparison (mean buoy winds)					
	2DVAR vs ECMWF	2DVAR vs buoy point wind	2DVAR vs ASCAT	Ν		
Default	1.91	2.01	1.22	6908		
New	2.06	1.85	0.81			

Data volume 15-03-2008

▶1 424 147 observations

Improved prediction of landing times by ModeS aircraft winds

Case\ Par- ameter	Minimum (s)	Maximum (s)	Mean (s)	St.Dev. (s)
No Wind	-293	169	2,3	79,9
KNMI 1.0	-80	70	-3,8	20,5
D11	-64	56	-3,2	17,7
H11	-58	46	-3,3	17,6
M11(3)	-69	55	-3,4	17,7
M11(1)	-61	50	-4,9	17,4

ModeS winds have impact in HIRLAM, but not in HARMONIE ?

HARMONIE model (Hirlam ALADIN Research on Meso-scale Operational NWP in Euromed)

- Non-hydrostatic
- ➢ 800x800 grid
- 2.5 km grid size
- 65 vertical levels
- 3D-Var assimilation
 - 8 times per day
 - 48-hour forecast
- ECMWF boundaries
- Available since 2012

HARMONIE impact experiments

- ➢ 6-week period 15/11/2013 31/12/2013
 - Including 5/6 December "Mandela storm"
- 3D-Var data assimilation
- Conventional observing systems:
 - radiosonde, aircraft, SYNOP (ground stations), buoys
- Available scatterometers: ASCAT A/B (12.5 km coastal), OSCAT (50 km), HSCAT (so far used for verification only)
- Experimental model version; cycle Cy38h1.2

NO OBS	; no observations
CONV-3h	; conventional observations – 3-h cycling
CONV+SCAT-3h	; CONV-3h + scat observations - no thinning
CONV+SCAT-THINN-3h	; CONV+SCAT-3h but ASCAT thinning (100 km)
CONV+SCAT-THINN-1h	; CONV+SCAT-THINN-3h but 1-h cycling

SCAT impact largely gone after 3 hours

- Possible explanations
 - Incorrect weight given to observations in the analysis; this was verified and indeed too much weight is currently given to observations.
 - For ASCAT: 1.39/1.55 stddev, ignoring "footprint error"
 - Timing issue in 3D-Var for e.g. aircraft, all satellite data.
 - But scores do not take into account coarse temporal data coverage
 - Most forecast initial states had no SCAT data
 - Probably better to limit verification to forecasts initiated with SCAT
 - Model bias (next slides)

Model bias example: Storm Ulli, **3 January 2012**

3 Jan. 2012 ~ 13UTC. In the strong westerly flow, a cold front rapidly moved across the North Sea, passing the Dutch coast. The front was accompanied with a squall line. The Dutch coastguard reported a so called meteo-tsunami at the coast of Ijmuiden, with a sea level change (rise and fall) of over 1.5 meters in 30 minutes.

EC-U10; OPER; verification time: 2012010314UTC

Harmonie 3 January around 13 UTC

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Harmonie 3 January around 14 UTC

Large impact SCAT on analyses but not maintained in a biased model

Conclusions

- > Mesoscale data assimilation is a new paradigm
- Many accurate 4D wind observations are needed to initialize 3D turbulence and convection in the atmosphere
- Undetermined scales cause headaches and destroy the analysis of the larger scales potentially
- It is possible to determine small observed scales in the analysis, if they did not exist yet (2DVAR)
- Weather models return to their climatological balance very quickly though
- Seek ways to avoid analyzing non-deterministic scales and their detriment as model noise
- Accurate treatment of time and space aspects, balance

Workshop

Wind Profiles and Mesoscale Data Assimilation

Ljubljana, 19-20 September 2016 meteo.fmf.uni-lj.si/en/workshop

Satellite Wind Services at Sea

- Asia India **OSI SAF** m Ocean
- 24/7 Wind product services (OSI SAF)
 - Constellation of satellites
 - High quality winds, QC
 - Timeliness 30 min. 2 hours
 - Service messages
 - QA, monitoring
 - Software services (NWP SAF)
 - Portable Wind Processors
 - Weather model comparison
 - Organisations involved: KNMI, EUMETSAT, EU, ESA, NASA, NOAA, ISRO, SOA, WMO, CEOS, ..
 - Users: NHC, JTWC, ECMWF, NOAA, NASA, NRL, BoM, UK MetO, M.France, DWD, CMA, JMA, CPTEC, NCAR, NL, . . .

More information:

www.knmi.nl/scatterometer

Typhoon Chan-hom, July 3, 2015 (early stage)

Typhoon Chan-hom, July 3, 2015 (early stage)

Typhoon Chan-hom, 2DVAR analyses

Typhoon Chan-hom selections

Wind front

Wind front 2DVAR analysis

Wind front selections

Triple collocation

	Scatterometer		Buoys		ECMWF	
m/s	σ_u	σ_{v}	σ _u	σ_v	σ_u	σ_{v}
ASCAT-A 25-km	0.63	0.71	1.21	1.35	1.39	1.44
ASCAT-B 25-km	0.63	0.66	1.26	1.39	1.38	1.42
ASCAT-A Coastal	0.76	0.84	1.18	1.34	1.54	1.57
ASCAT-B Coastal	0.81	0.79	1.24	1.35	1.53	1.57

QC: Which error is acceptable?

- We can produce winds with SD of buoy-scatterometer difference of 0.6 m/s, but would exclude all high-wind and dynamic air-sea interaction areas
- The winds that we reject right now in convective tropical areas are noisy (SD=1.84 m/s), but generally not outliers!
- What metric makes sense for QC trade-off?

Observations and Models

T2m verification over land

NO OBS CONV-3h CONV+SCAT-3h CONV+SCAT-THINN-3h CONV+SCAT-THINN-1h

- Scatterometer improves analyses of 2-m temperature over
- land

 \geq

cases

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- SCAT impact gone after 3 hrs (3-h cycle)
- Impact maintained for 1-h cycling; SCAT or additional SYNOPS??

NO OBS CONV-3h CONV+SCAT-3h CONV+SCAT-THINN-3h CONV+SCAT-THINN-1h

Scatterometer slightly positive for 10-m wind over land Similar sores for SCAT thinning/nothinning

Tandem-Aeolus impact on **analyses** Analysis improvement at forecast initial time of '99

Christmas storm Martin (26 Dec 1999 12:00 UTC) for the

Tandem-Aeolus scenario Single-time SOSE; 6 hours DWL obs.

ANALYSIS IMPROVEMENT (m/s) 500 hPa (u,v); andate=19991226 12UTC

SOSE – cycling; 84 hours DWL obs.

N.HEMIS: 0.54. EUR: 0.03. NATL: 0.25. NAMER: 0.13. N.PAC: 1.02. N.ASIA: 0.37. N.POLE: 0.72 10 2.5 0.5 -0.5 -1 -2.5 -5 -10 25

> Positive interference of subsequent cycles

All the QC-accepted and 2-solution (|MLE₁|<1)

	ASCAT-ECMWF-buoy comparison (mean buoy winds)					
	ASCAT vs ECMWF	ASCAT vs buoy point wind	Ν			
Default	2.19	1.74	5034			
New	2.17	1.71				

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	ASCAT-ECMWF-buoy comparison (mean buoy winds)				
	2DVAR vs ECMWF	2DVAR vs buoy point wind	2DVAR vs ASCAT	Ν	
Default	1.85	1.94	1.17	5034	
New	2.00	1.76	0.74		