



Norwegian
Meteorological
Institute



Highlight of recent progress in HIRLAM upper-air data assimilation

Roger Randriamampianina
with contribution from HIRLAM colleagues

EWGLAM, 2018, Salzburg, Austria

outline



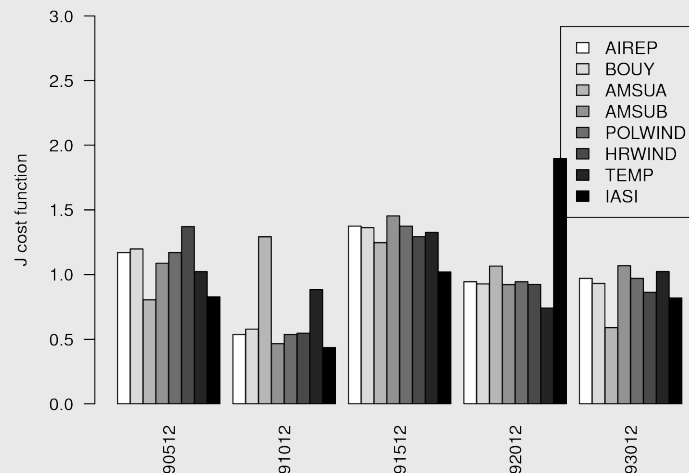
- Operational upper air data assimilation (UA-DA) systems in HIRLAM
- Algorithmic development
- Improving the use of observations
- Concluding remarks and further plan

Operational upper air data assimilation (UA-DA) systems



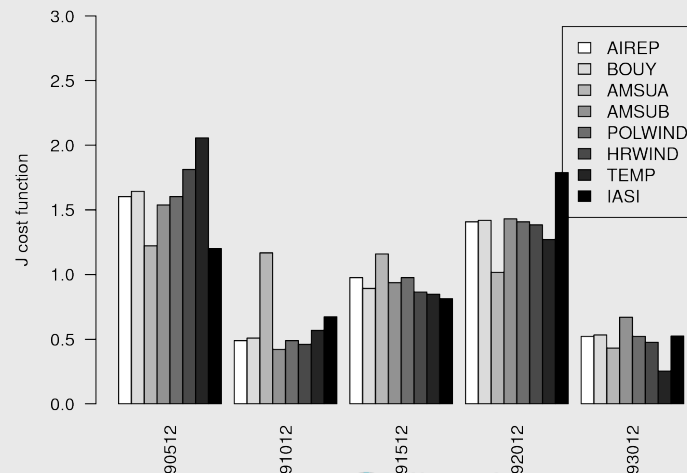
- **Assimilation scheme:** 3D-VAR;
- **Cycling Strategy:** 3 hourly;
- **Conventional observations:** SYNOP, SHIP, BUOY, AMDAR, AIREP, ACARS, ModeS EHS, Pilots, TEMP;
- **Satellite radiances:** AMSU-A, AMSU-B/MHS, ATMS, IASI;
- **Satellite retrievals:** Scatterometer, GNSS ZTD, GPS RO, **AMV**;
- **Radar observations:** Reflectivity;
- **Bias correction scheme:** Variational (VarBC).

Normalized variability of the cost function over different dates
Forecast: 6 hours, Total Norm



Experiment run
with
AROME-MetCoOp
Sensitivity of the
forecast model to
different
observations

Normalised variability of the cost function over different dates
Forecast: 48 hours, Total Norm



Field Alignment (FA) & Variational Constraint (VC)

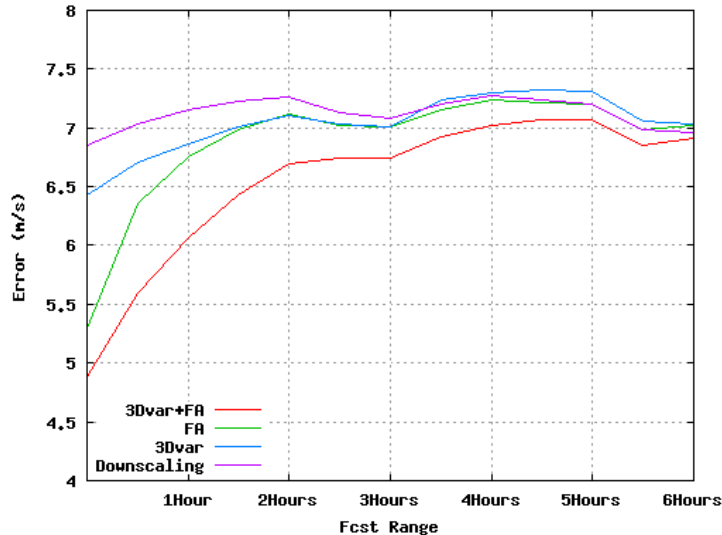
Carlos Geijo

Assimilation of Doppler Wind Radar Data in HARMONIE

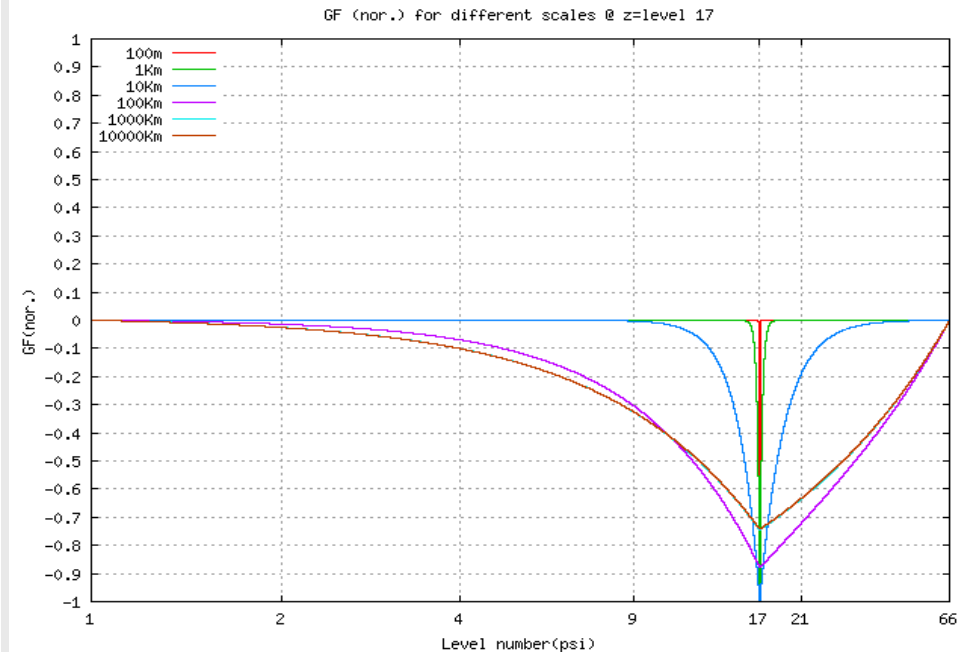
- Verification of forecasted radial wind using the own radar data:

$$\text{Error} \equiv \sqrt{\langle (\text{Fcst} - \text{Radar})^2 \rangle_{\text{PPI}=0.5}^{1/2}} + \sqrt{\langle (\text{Fcst} - \text{Radar})^2 \rangle_{\text{PPI}=1.4}^{1/2}}$$

- Results averaged over more than 150 cases:



Testing the FA in HARMONIE system

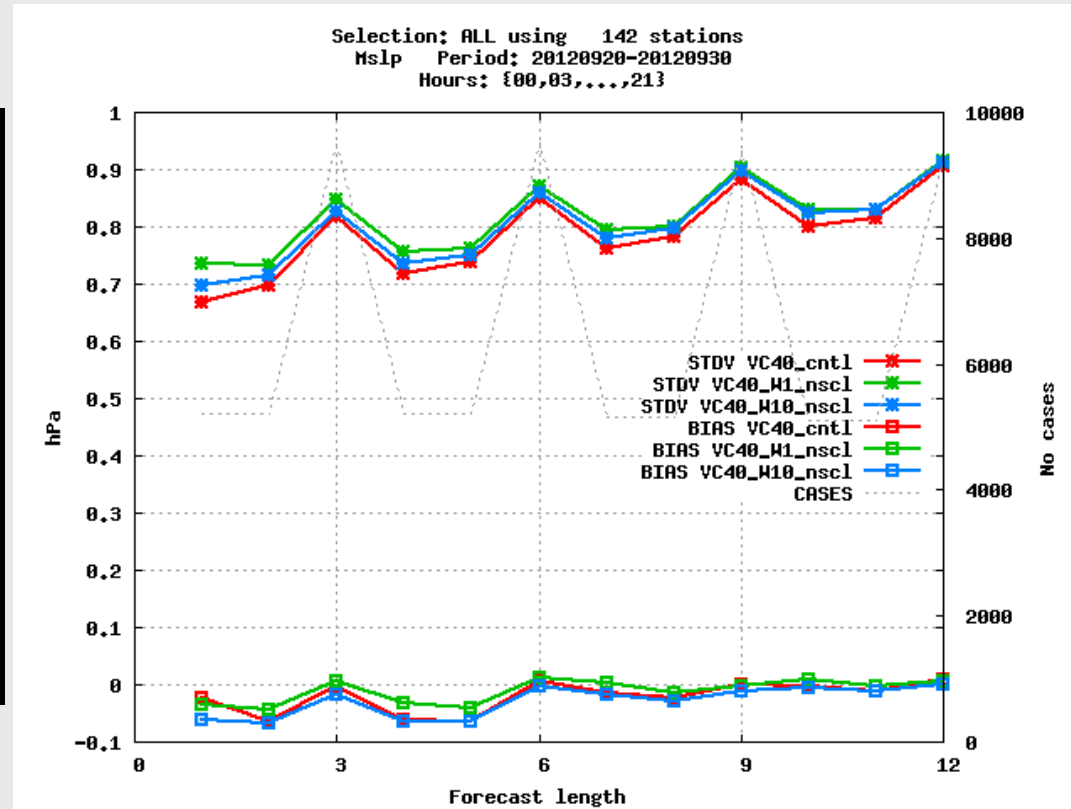
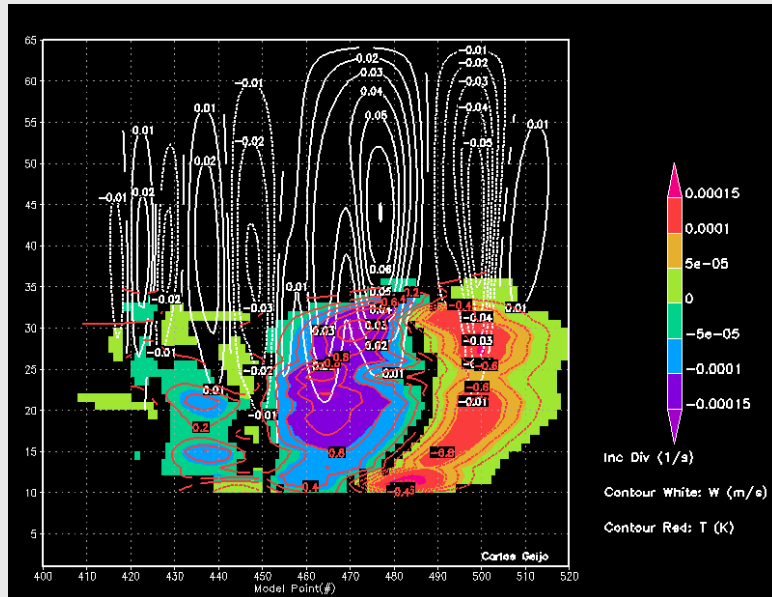


Property of the Green Function

FA and VC

Carlos Geijo

Balanced Vertical Motions



Still issue with horizontal spreading/
balance

3h-DA cycles with conventional obs ($\sim 3 \cdot 10^3$ obs/cycle) . No surface analysis.

Control: 3D-VAR (Red line)

Experiments: (Green ($w=1$) and blue ($w=10$) lines) 3D-VAR with no statbal (LUNIVARIATE=.T.) and VC for balancing. Verification using the standard HIRLAM tool.

Ensemble nowcasting

Xiaohua Yang

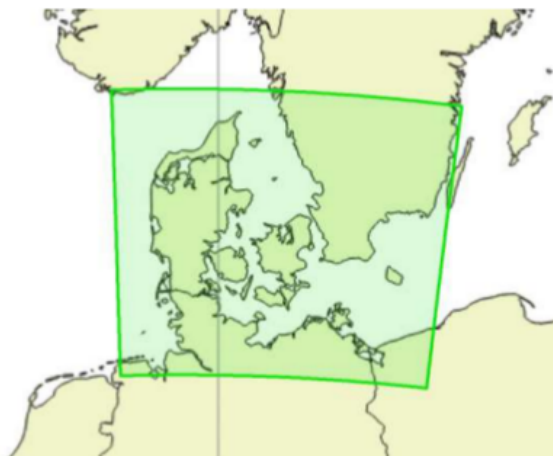
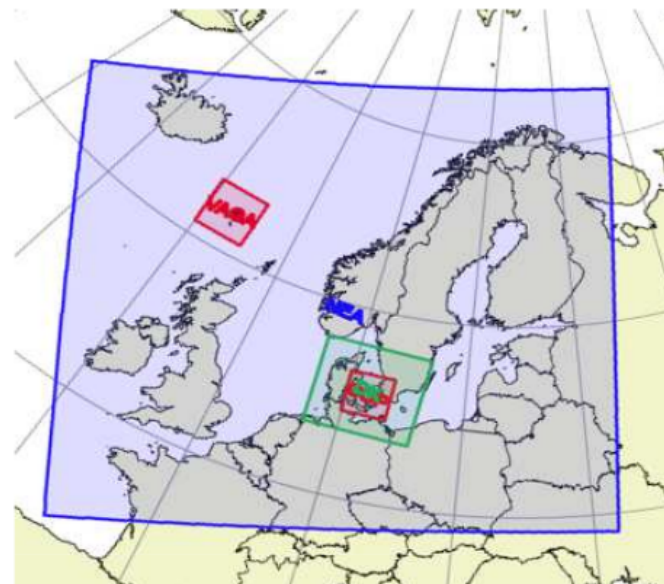
Operational COMEPS@DMI

24 +1 members, Harmonie-arome@2.5 km

1200x1080x65 gridmesh

Time-lagged EDA/EPS with

- hourly 3DVAR (control)
- 4 perturbed members each hour, 57h forecast
- Perturbation (EDA)
 - observation, multi-physics, surface, SLAF



COMEPS-nowcasting (prototype)

Basics: Ensemble Harmonie-arome @750m + radar extrapolation, 7 to 13 NWP members,, 800x720x65
10 min update with time-lagged EDA/EPS

- 3DVAR every 10 min, 9 h forecast
- COMEPS@2.5 km as LBC

Upper air DA – Ensemble nowcasting – Xiaohua Yang



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Toward a nowcasting RUC at sub-km

Motivation: DMI wishes to develop a forecast capability to warn rapidly developing convection which often is characterised by a **very short life cycle (1-3 h)**, **very small scales (few km)**.

Approach: an ensemble RUC with NWP-based sub-km resolution nowcasting + radar extrapolation (in first hour)

- **Model:** sub-km Harmonie-arome with capability for small scale convection; possibly + nudging
- **Assimilation:** variational approach with 3DVAR/4DVAR
- **Ensemble:** a time lagged EDA with RUC on overlapping windows

Extension of DMI-COMEPS from hourly, 2.5 km system toward sub-hourly and sub-km

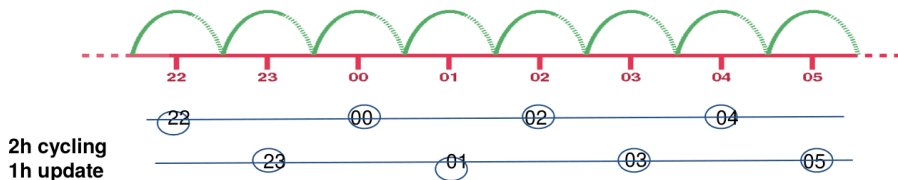
Also use of more observations from crowd can be considered



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How to achieve frequent analysis update

Cycling strategy with intermittent DA



Assimilation window **2h**, data between -1 and 1 h
Background from **2h** forecast of previous cycle

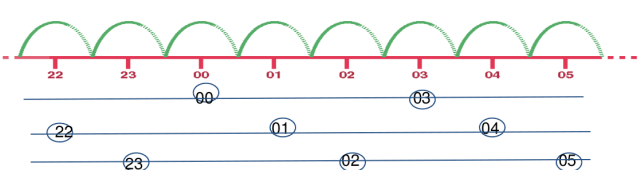


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How to achieve frequent analysis update

Cycling strategy with intermittent DA

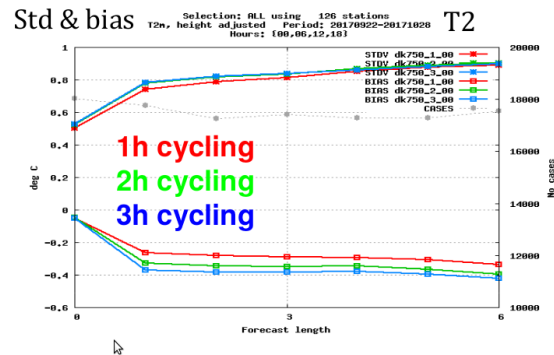
3h cycling
1h update



- + extended use of obs (longer window, less thinning)
- + reduced spin-up
- + feasible for larger, time lagged ensemble
- + better load balance for ensemble model

Assimilation window **3h**, (-1.5 and 1.5 h)
Background from **3h** forecast of previous cycle

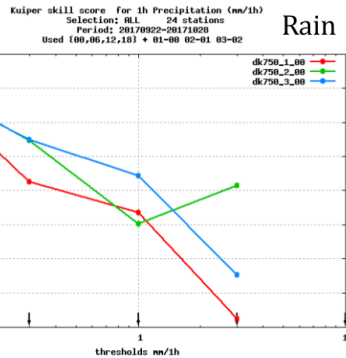
Upper air DA – Ensemble nowcasting – Xiaohua Yang



Dilemma about assimilation window length & cycling frequency

1h cycling clearly advantageous for dry surface parameters like MSLP/T2m
But for cloud and precipitation, 1h cycling is less optimal compared to 2h/3h

The situation may depend on obs data, analysis schemes, ...



Independent, parallel
cycling system

Results
according to today's model accuracy!

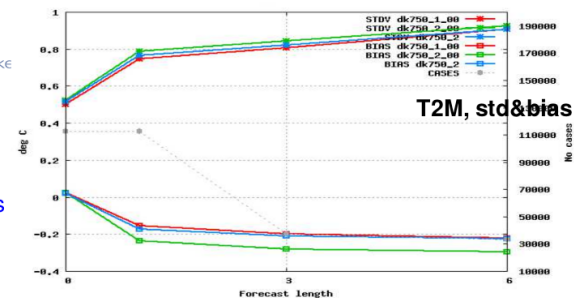
Danish Meteorological Institute

Xiaohua Yang, EUMETNET Workshop on Ensemble-Nowcasting, 16-18 May 2018

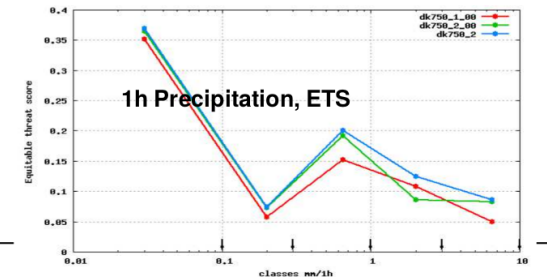
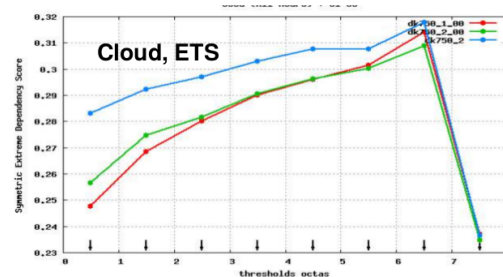
partially connected system



Partial coupling
between adjacent suites



Reducing the moisture spinup!



4D-Var development

Nils Gustafsson, Jan Barkmeijer, Magnus Lindskog, Jelena Bojarova

HARMONIE Multi-incremental 4D-Var – some technical improvements

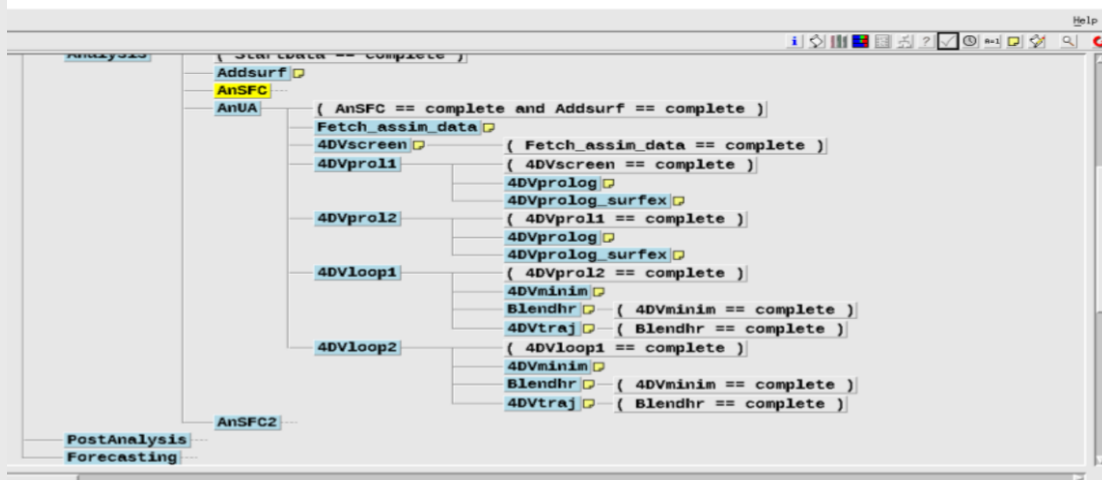
- Multi-incremental 4DVAR with multiple resolutions
(different resolutions in different outer loops)
- 2D parallelization
- Treat the humidity in spectral space
- Increment for TL model = 0 at first iteration
- Handling of VARBC coefficients in control vector file
- Problem with skin temperature for AMSU assimilation in second outer loop
 - “fixed” by using model skin temperature. Using skin temperature and emissivity as “slack variables” during minimization could be considered.
- “EZONE” has to be increased in case of very coarse resolution outer loops

- Tested with shorter period runs with “operational observations”

4D-Var development

Nils Gustafsson, Jan Barkmeijer, Magnus Lindskog, Jelena Bojarova

HARMONIE Multi-incremental 4D-Var – affordable operationally



Outer 4dVar loops with different horizontal resolution in different outer loops

bg

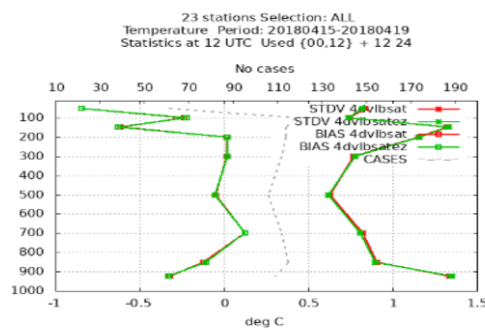
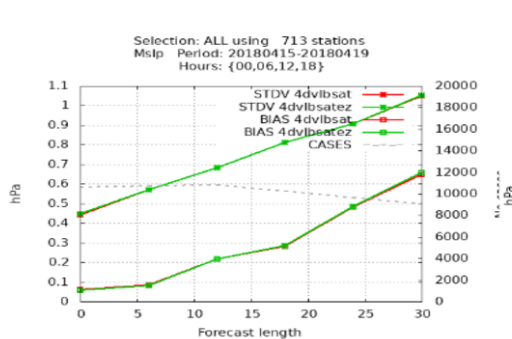
fg

Control vector in spectral space – padded with zeroes when resolution is increased

4dvlbsat : 30 iter 3 dx + 30 iter 3 dx

4dvlbsatez : 20 iter 6 dx (needed to extend EZONE to 25) + 10 iter 3 dx

Both with satellite and conv. data.



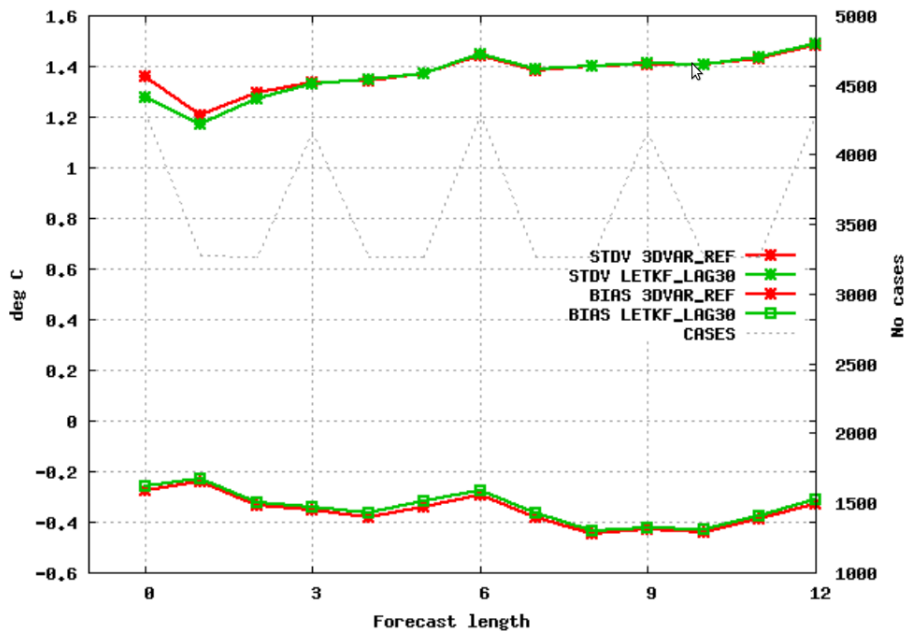
LETKF

Pau Escriba & Jelena Bojarova

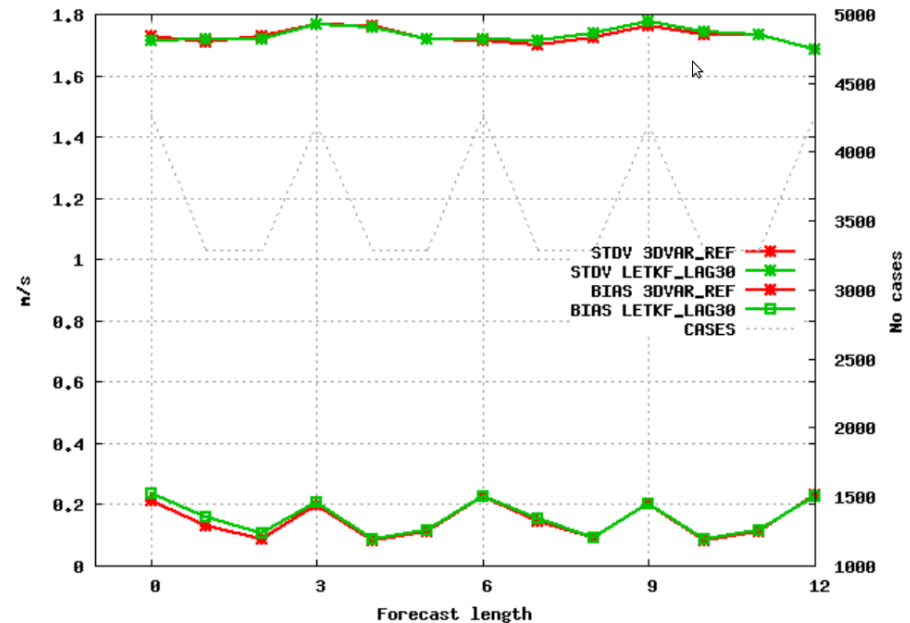
- LETKF is quite mature after several tunings but still not finished
- In a 8 day (short) verification period **the LETKF performance is comparable to 3DVAR, even better in some parameters when assimilating only conventional observations**
- Many short-period testing have been done
- More verification with longer periods is needed

More about the tests and the results can be found on our last working week page

Selection: ALL using 163 stations
T2m Period: 20170401-20170408
Hours: {00,06,12,18}



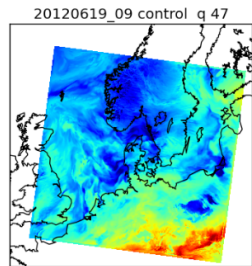
Selection: ALL using 161 stations
U10m Period: 20170401-20170408
Hours: {00,06,12,18}



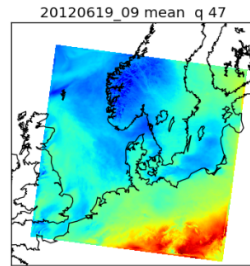
Hybrid EnVar implementation and Brand

Brand: A perturbation technique through B

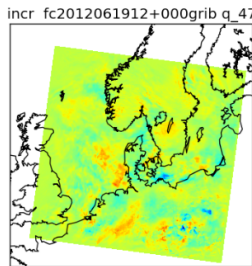
Jelena Bojarova



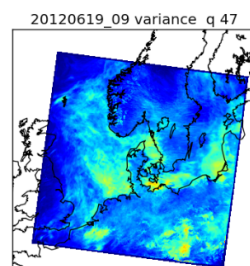
First Guess HybridEnVar



Mean BRAND "after" (+03h)



Analysis Increment



STD BRAND "after" (+03h)

				Hybrid EnVar (BRAND “after” Small spread)		Hybrid EnVar (BRAND “after” Large spread)		
				#obs	“o-b”	“o-a”	“o-b”	“o-a”
SYNOP	LAND	Z	87	18.03	4.61	15.53	2.23	
		Z	297	47.98	12.44	53.56	6.39	
	SHIP	Z	16	2.49	1.22	7.06	0.82	
		U	10	8.84	6.16	8.25	3.82	
		Z	28	10.37	6.31	11.27	2.95	
		U	16	13.08	9.11	13.79	4.59	
AIREP	AMDAR	U	124	37.86	11.99	44.28	4.28	
		T	62	10.36	6.07	13.08	2.32	
DRIBU		Z	1	0.01	0.02	0.07	0.01	
TEMP		U	2362	960.48	669.28	1098.07	629.02	
		T	613	618.36	485.95	648.85	463.03	
		Q	345	366.04	126.73	478.10	114.29	
J		Jo	3961	2093.96	1339.93	2391.96	1233.79	
		Jb			120.42		82.31	
		Je			60.85		84.85	

20120613 00 UTC

EnVar: implementation following *Gustafsson and Bojarova (2014)* is done and testing and further development of the scheme is ongoing.

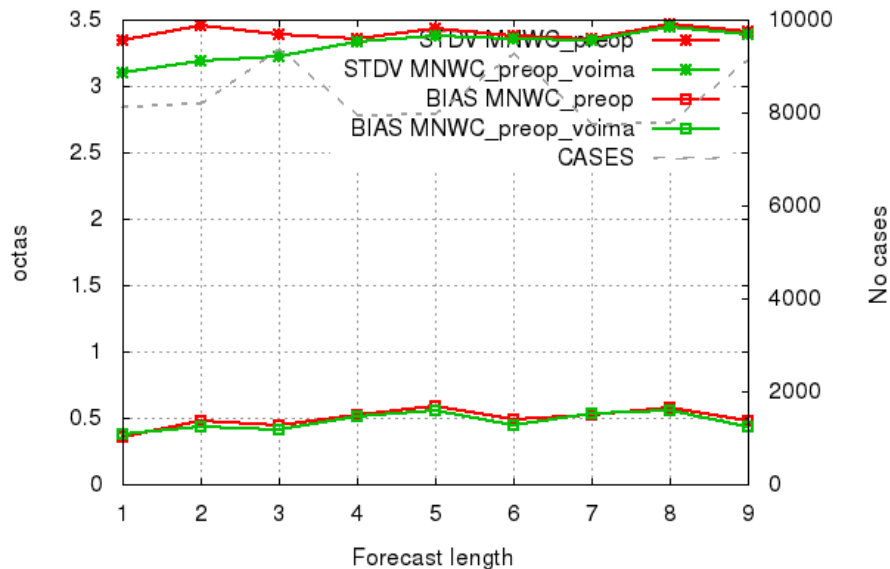
Brand: Implementation is finished. Tuning of the technique is needed.

Using cloud data in nowcasting initialisation

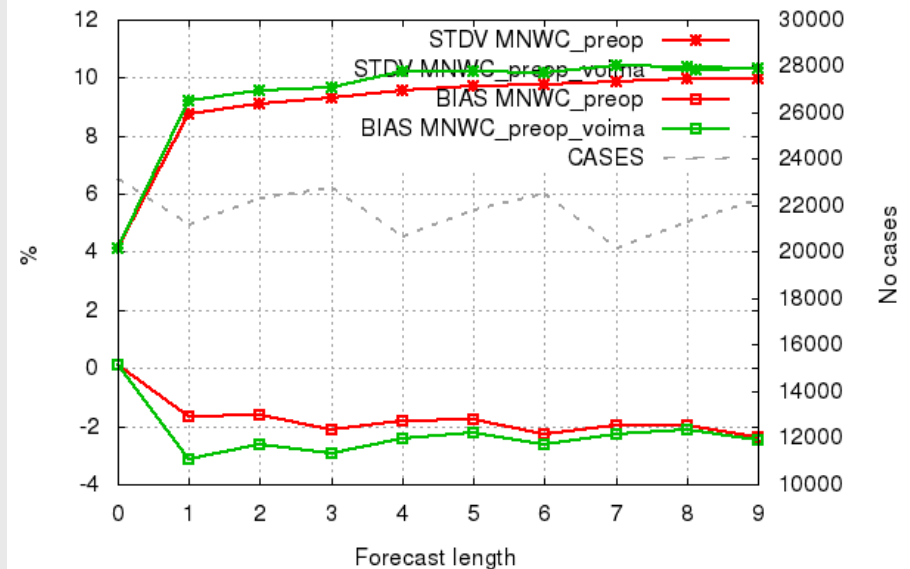
Erik Gregow

- 7-days: 29 Aug- 4 Sep 2018
 - **MNCW_preop** – rapid refresh without MSG cloud initialisation
 - **MNCW_preop_voima** – rapid refresh with MSG cloud initialisation
- Good impact on cloudiness, but further tuning is needed for the other verified parameters.

Selection: ALL using 284 stations
Cloud cover Period: 20180829-20180904
Hours: {00,03,...,21}



Selection: ALL using 606 stations
Rh2m Period: 20180829-20180904
Hours: {00,03,...,21}



Variational bias correction for aircraft data

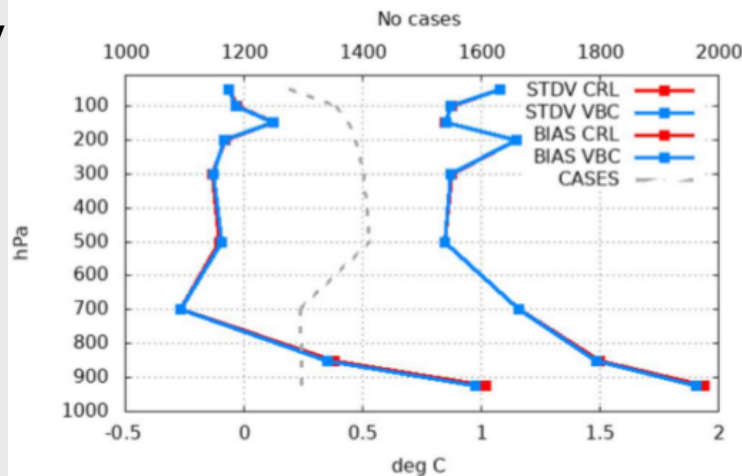
Magnus Lindskog

- Test with the off-set only;
- Seems to work well and do the bias corr.;
- Promising positive impact;
- Issue with growing file size;

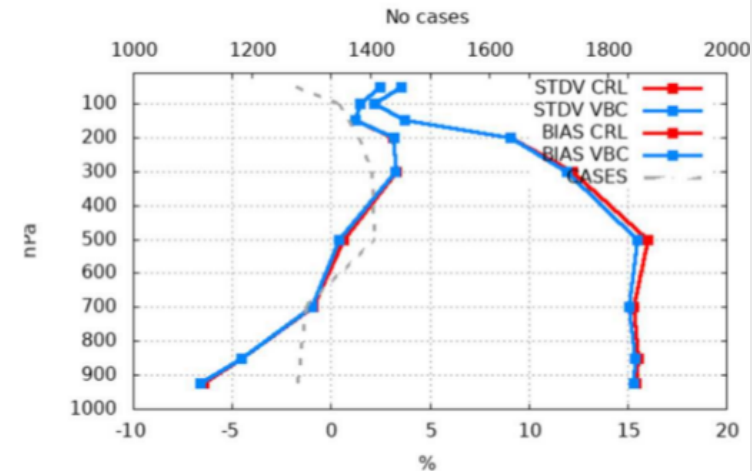
Verification scores

SMHI

12 stations Selection: ALL
Temperature Period: 20170101-20170204
Used {00,12} + 12 24



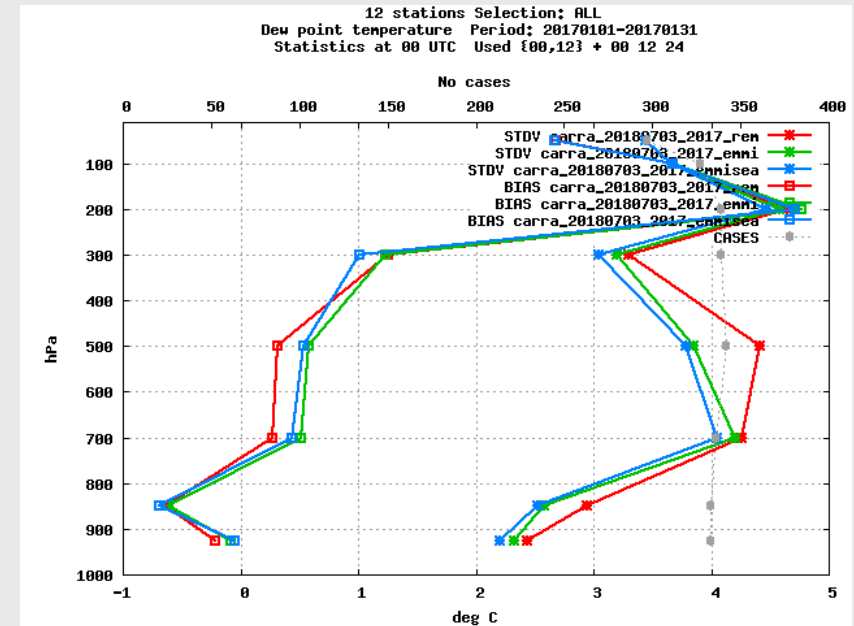
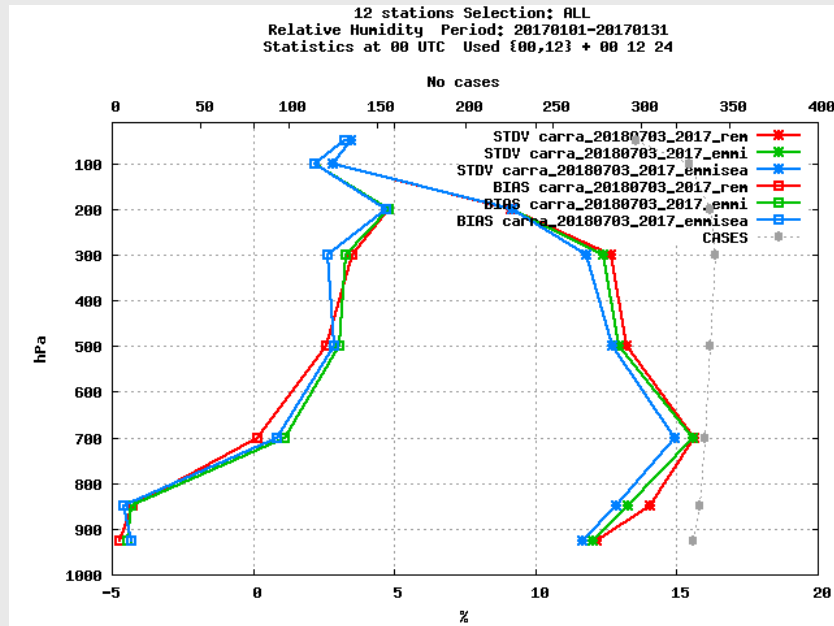
12 stations Selection: ALL
Relative Humidity Period: 20170101-20170204
Used {00,12} + 12 24



Neutral to slightly
positive impact

Use of emissivity Atlas in microwave radiance assimilation

Sigurdur Thorsteinsson & Roger Randriamampianina



Very promising results!

Blue – use of emissivity over sea only;
Green – use of emissivity over sea and land;
Red – Not using the emissivity atlas.

More on the use of observations

- Implementation of GNSS slant delay assimilation with very promising impact.
- MET Norway is involved in the CAL/VAL project with the Aeolus HLOS wind data. Our newest system 43h2 is ready for this challenge.
- Máté Mile is starting PhD with, among other tasks, special focus to effective resolution of observations (accounting for the real footprint of observations) in DA.
- We think it is very important to make all verified (“assimilable”) observations available in the operational systems. Specific resource is dedicated to this task at different centres.
- We are involved in Copernicus reanalysis for both European and Arctic regions.
 - allow short development works like improving the observation pre-processing, optimisation of the DA system, improvement of blacklisting procedure.

Concluding remarks and outlook

- Continue the local implementation of more observations ...;
- Continue exploring 1-h DA update (overlapping windows) and Rapid refresh;
- Working with initialisation schemes: LHN, use of variational constraint, IAU;
- Continue developing the 4D-VAR, LETKF and hybrid EnVar schemes;
- Further improve the treatment of radar data;
- **Improve the observations pre-processing:** Bator for all observations and at the same time develop COPE to handle all observations;
- **Diagnose B computation** by checking Hirlam and MF/Aladin ways of computation;
- **Better accounting of large scale information** in initialisation and data assimilation;
-

Thank you