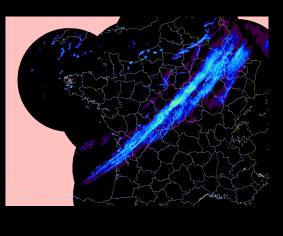
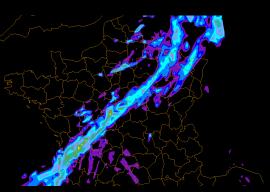
Implementation of the 1D+3DVar assimilation of radar reflectivities in the AROME model at Météo-France











Eric Wattrelot



Outlines

- Introduction
- Evolution of the radar product for AROME
- Principle of the 1D+3DVar method of reflectivities assimilation
- Illustration of the 1D method, problematic of assimilating the "norain" signal
- Evaluation of the 1D method (towards the quality control for the following 3DVar)
- Some few results of the 1D+3DVar through case studies: distinction between stratiform and convective rain cases
- Objective evaluation of two months of cycled assimilations in the 3hour Rapid Update Cycle AROME
- Conclusions and perspectives



Context and Introduction

Problematic of the assimilation method of reflectivities:

- •Reflectivity observation operator needs a complete description of warm and cold hydrometeors: realistic simulation can be obtained with AROME.
- •But if reflectivities can provide useful information about the atmosphere water cycle (rain, snow, graupel, primary ice), in the context of variational assimilation, assimilation of rain is very difficult, because:
 - ✓ The direct observation operator involves physical processes which are characterized by discontinuities and nonlinearities, and there is need of simplification in the linearized versions to get some good results...
 - ✓ "rain" is not a variable which is in the "control variable" of the analysis
- But rainfalls have a short shelf life in the atmosphere. Therefore, it's better to try to modify only the humidity field >> need of a 1D method to get some relative humidity retrievals from reflectivities (following ideas of Mahfouf, Marecal ...)

Status of the assimilation of radar reflectivities in AROME:

- •Caumont, 2006 has introduced the 1D baysian inversion of reflectivities (offline in Meso-NH)
- In the 3DVar (RUC) AROME, many sensibilities studies, impact studies on specific cases, adjustments and tuning of the 1D+3DVar assimilation method of reflectivities and long period currently being tested in a pre-operational framework (Wattrelot & al., 2008)



The AROME product from the **French Aramis network**

24 radars: 16 in C band (yellow circles) + 8 in S band (green circles). Volumes reflectivities (from 2 to 13 elevations).

Pluie

Bruine.

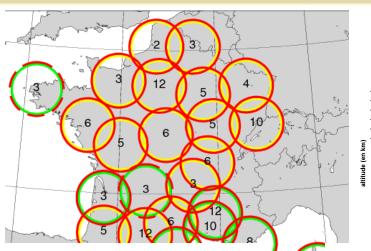
Ciel clair.

Bruit

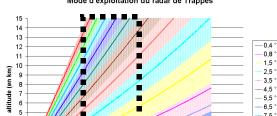
22 Doppler radars

Rain **Drizzle** Sea clutter **Clear sky** echoe Ground clutter

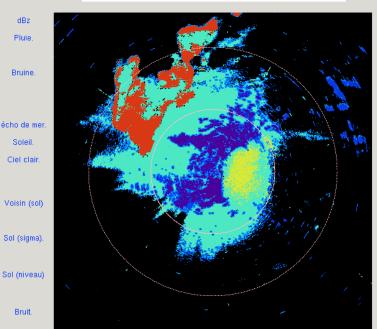
Noise



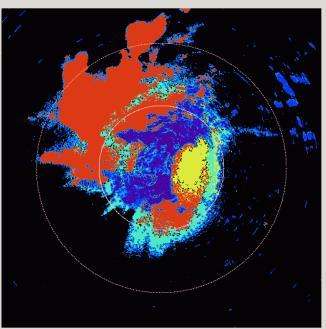
Optimal vertical sample at medium range
Mode d'exploitation du radar de Trappes



OLD PRODUCT



NEW PRODUCT



Radar reflectivities assimilation: inversion method

The α best α estimate of atmosphere α given the observation α 0 and using Bayses's theorem (Lorenc, 1986)

$$E(x) = \iiint x.P(x = x_{true} \mid y = y_0)dx$$

$$E(x) = \iiint x P(y = y_0 \mid x = x_{true}) P(x = x_{true}) dx$$

Olson, 1996 (Gaussian and uncorrelated errors) and x_j database of atmospheric profiles

$$E(x) = \sum_{j} x_{j} \frac{\exp \frac{-1}{2} \cdot || y_{0} - y_{s}(x_{j}) ||^{2}}{\sum_{j} \exp \frac{-1}{2} \cdot || y_{0} - y_{s}(x_{j}) ||^{2}}$$

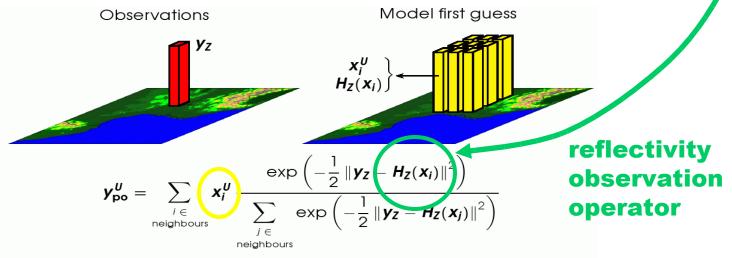
(with
$$||y_0 - y_s(x_j)||^2 = [y_0 - y_s(x_j)]^T \cdot (O + S)^{-1} \cdot [y_0 - y_s(x_j)]$$
)



Inversion method of reflectivities profiles

Caumont, 2006: use of model profiles in the vicinity of the observation as representative database

$$E(x) = \sum_{j} (x_{j}) \frac{\exp \frac{-1}{2} \cdot || y_{0} - (y_{s}(x_{j}))||^{2}}{\sum_{j} \exp \frac{-1}{2} \cdot || y_{0} - y_{s}(x_{j})||^{2}}$$



 y_{po}^{U} : column of pseudo-observed relative humidity,

 $\dot{y_z}$: column of observed reflectivities,

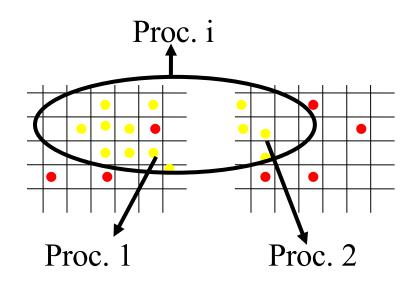
 $\mathbf{x}_{i}^{\mathbf{U}}$: column of relative humidity,

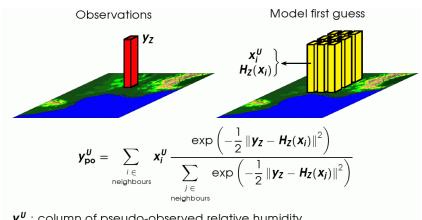
 $H_Z(x_i)$: column of simulated reflectivities.

- 1. Consistency between the retrieved profile and clouds/precipitations that the model is able to provide
- 2. But, possibility of bad solution if model too far from the reality



Challenge to implement the 1D inversion: use of the extension of the model/observation interface in 2 dimensions





Toujours un temps d'avance

 \mathbf{y}_{po}^{U} : column of pseudo-observed relative humidity,

yz: column of observed reflectivities,

 \mathbf{x}_{i}^{U} : column of relative humidity,

 $H_{Z}(x_{i})$: column of simulated reflectivities.

In context of Arpege/IFS code:

- Difficulty to use model profiles in the observation space: use of the semi-lagrangien interpolation for the profiles in the supporting database as it is done for observations
- and use of the new 2D-interface of model/observation built for 2D observation operator (Grace instrument from METOP).

Illustration with one radar assimilated, in the observation space **Arome (guess)** Radar -10-2 = 2-5 = 5-10 = 10-15 = 15-20 = 20-25 ■ 25 - 30 ■ 30 - 35 ■ 35 - 40 ■ 40 - 45 ■ 45 - 50 ■ 50 - 55 **25 - 30 30 - 35 35 - 40 40 - 45 45 - 50 50 - 55 Reflectivities Elevation** 0.44 20 - 40 = 40 - 45 **Relative humidity** Pseudo-obs. Arome (guess)

Toujours un temps d'avance

Problematic of « no-rain » s ignal

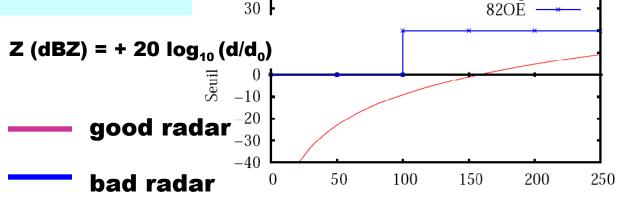
Problematic:

- if the SNR (signal-noise-ratio) is very low (bad quality of the radar or very far away from the radar), there is a high probability to dry with pixels below the noise but rainy!!
- ☐ But we know the noise, so we can ... take into account (in the 1D method) the threshold of detection for each pixel of each radar in order to not dry the model below this value : but is it sufficient?

Use only
$$ZZ_{SIM} > ZZ_{THR}$$
 and $(ZZ_{SIM} < ZZ_{THR} \Rightarrow ZZ_{SIM} = ZZ_{THR})$

☐ Sensibilities studies have been done by deterioration of the quality of one radar

Threshold of detection, function of range



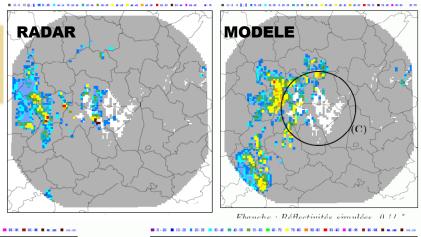
40



Courbe Theorique







good radar

bad radar

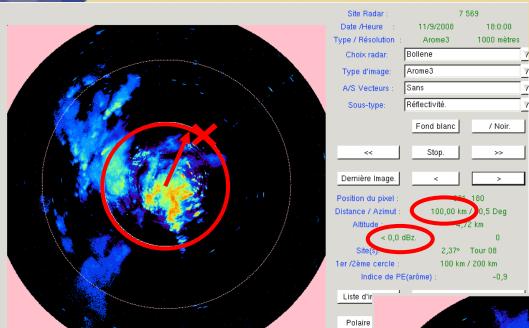
Solution: take into account the noise until values close to an acceptable value of "no-rain": example, 0 dbZ!



Wrong desaturation in slightly rainy areas (could happen if

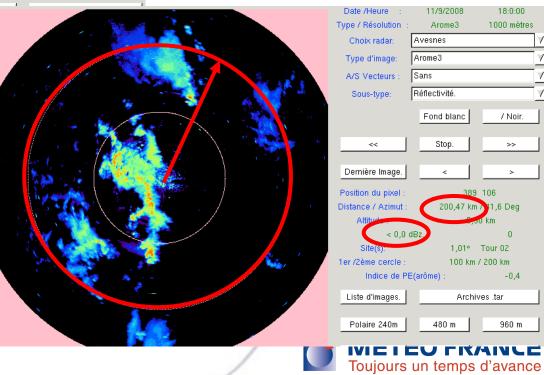
$$ZZ_{obs} < ZZ_{THR} < ZZ_{SIM}$$
)



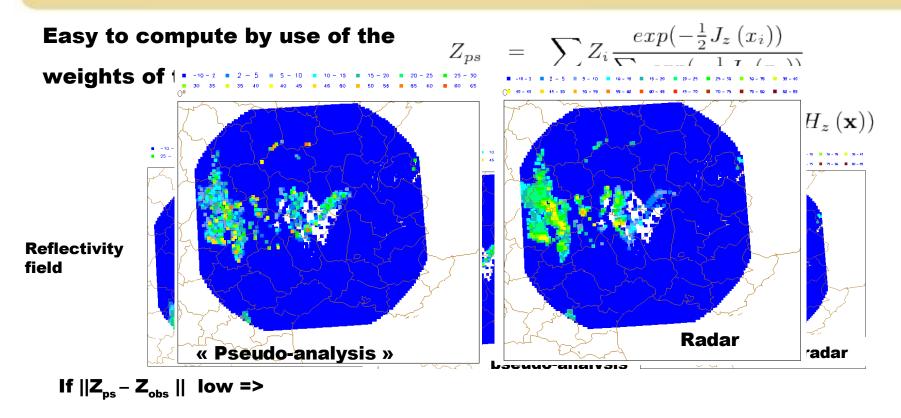


bad radar

good radar



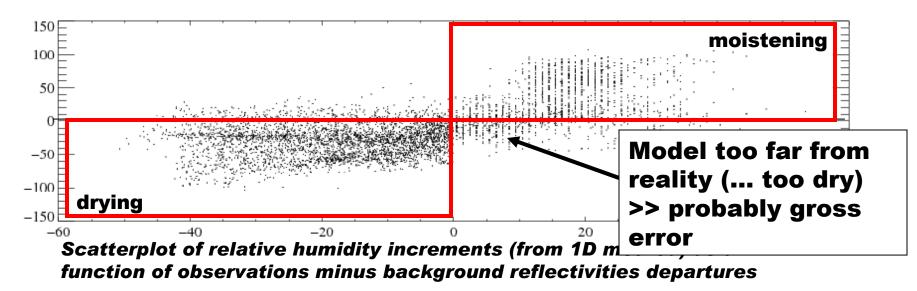
To evaluate the 1D-method, use of a « pseudo-analysis of reflectivity »



- 1. good convergence of the 1D method. (RMS deviation can be a measurement of the quality of the retrieved profiles: useful for monitoring)
- 2. good consistency between the pseudo-observations and the model (because of use of model information in the 1D-inversion). Used for the quality control in the screening
- 3. Possibility to take into account this value in the choice made in the thinning of the observations (tests are still underway)

Towards the Q uality control, thinning and errors...

Another diagnostic: check if drying and moistening are consistent with the negative or positive reflectivity departures: convergence involves that the model has the capability to saturate or desaturate close to the observation.



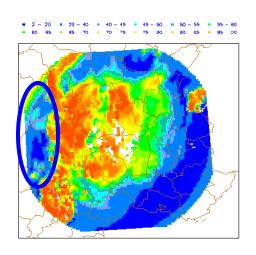
- thinning every 15 kms to avoid observation error correlation and representativeness errors (sensibilies studies have shown it is not useful to increase density of obs. with the system now)
- specification of pseudo-observation error variance, it depends linearly on the range from the radar until 160 kms, but low sensibility of the results to the observation error specification

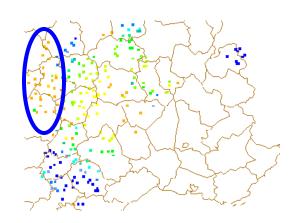


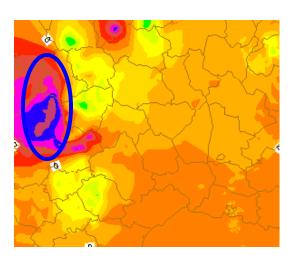
Pre-identification of « gross observation errors » allows to increase the acceptable value « obs mi nus guess departures »...

ACTIVE PSEUDO-OBSERVATIONS OF RELATIVE HUMIDITY

RELATIVE HUMIDITY ANALYSIS DIFFERENCES

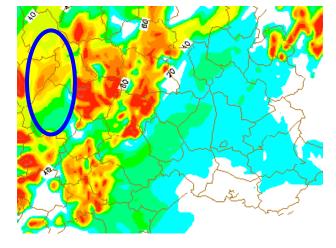




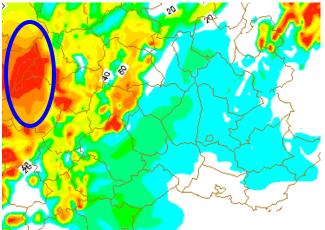


Toujours un temps d'avance

ANALYSIS
of relative
humidity
at 500 hpa



~30/50% obs minus guess departures



~50/70% obs minus guess departures METEO FRANCE

Case of stratiform rain

It works well in stratiform cases: generally because good spatial coverage of precipitation in the model. The method founds sufficient information to converge.

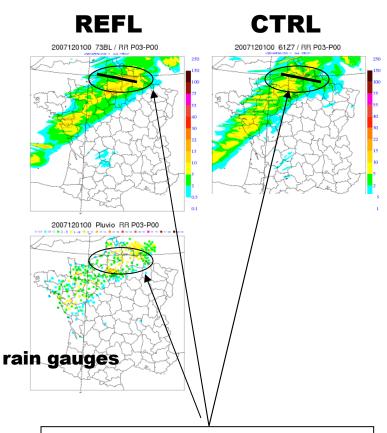
Illustration on one case...



Case of stratiform ra Modification of the divergence field in low band of cold fi

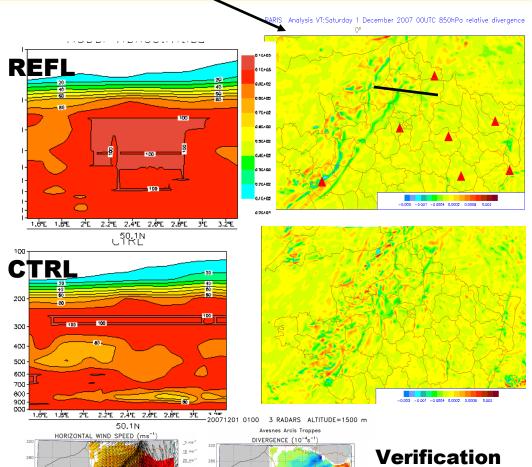
3h - cumulated rain - P3-P0

layer thanks to the cross-correlations of the B-matrix between humidity and wind field...



Capability to shift the cold front (well located on the 3-hour forecast from the analysis with reflectivities)

3 cycling



with

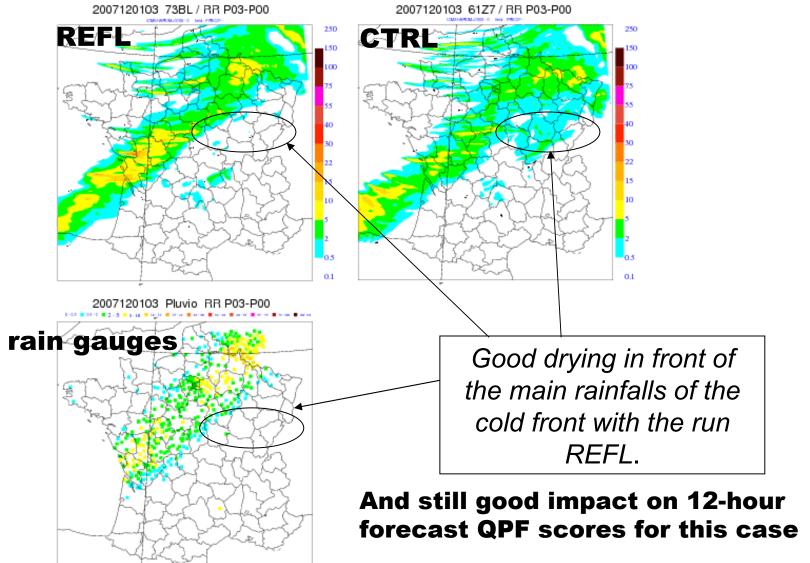
field

independant reconstruction

of the 3D wind

One cycling more: r03 - 4 cycling

3h - cumulated rain - P3-P0

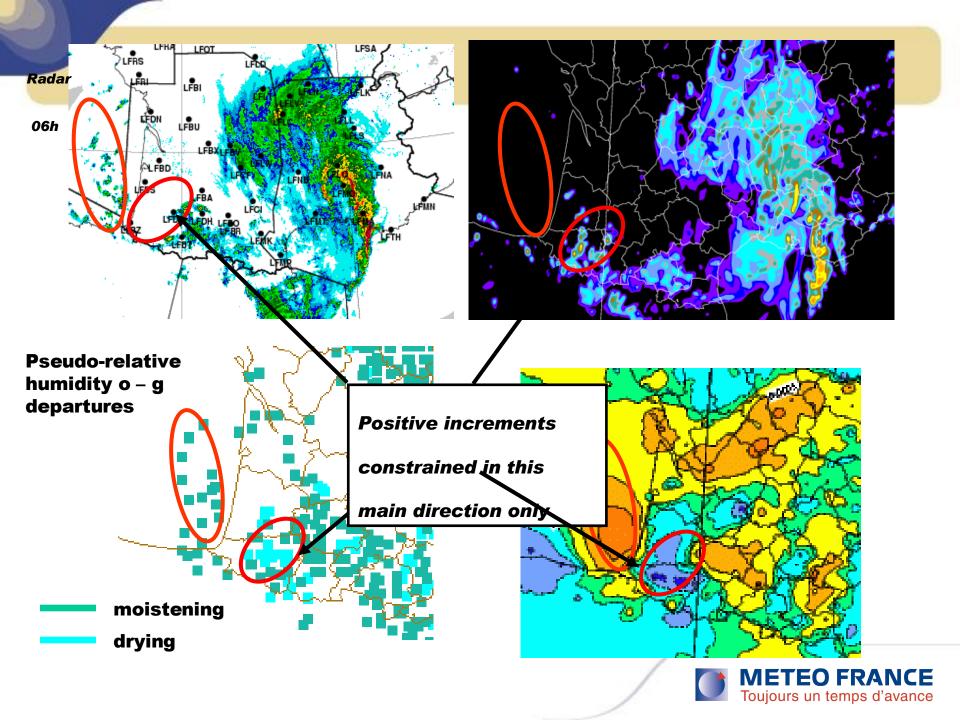


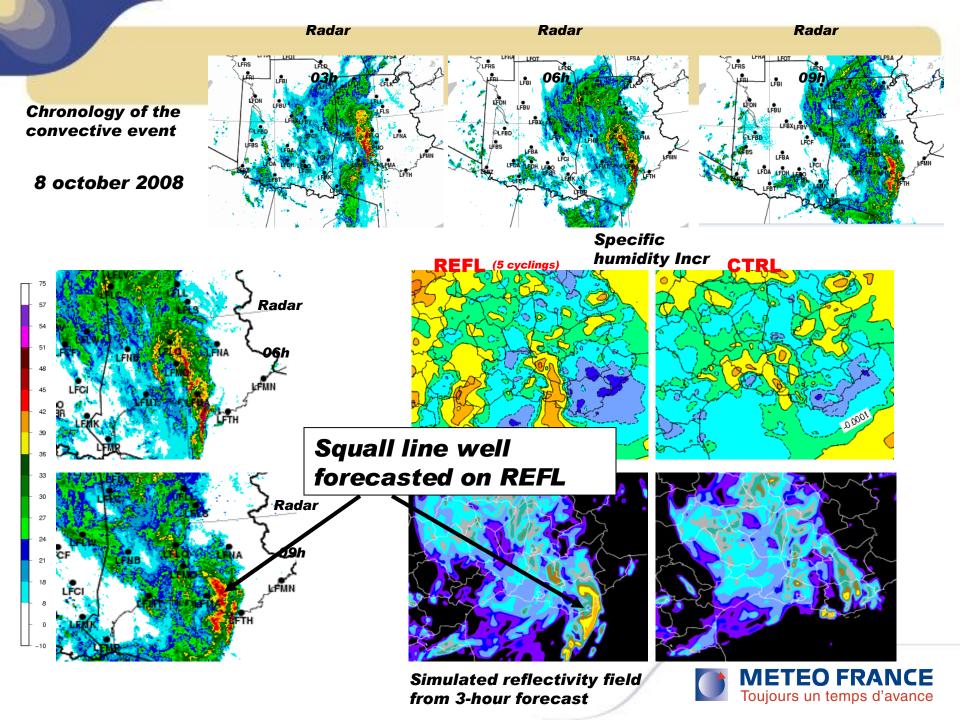


Case of convective rain: to simplify, what we have learnt...

- No convection triggered in the model (in the vicinity of the observation): difficulty to create precipitation by the 1D method (quality control in order to not introduce gross errors of pseudo-observations).
 - If convective cells exist in the model close to the observations: ok but...
- ■To avoid positive biais of humidity: use of observed "no-rain" pixels: ok to remove wrong cells and avoid spreading of positive increments
- If convective cells exist but very far from the observations: spreading of positive increments because of too big range length in B matrix... problem needed to be evaluated



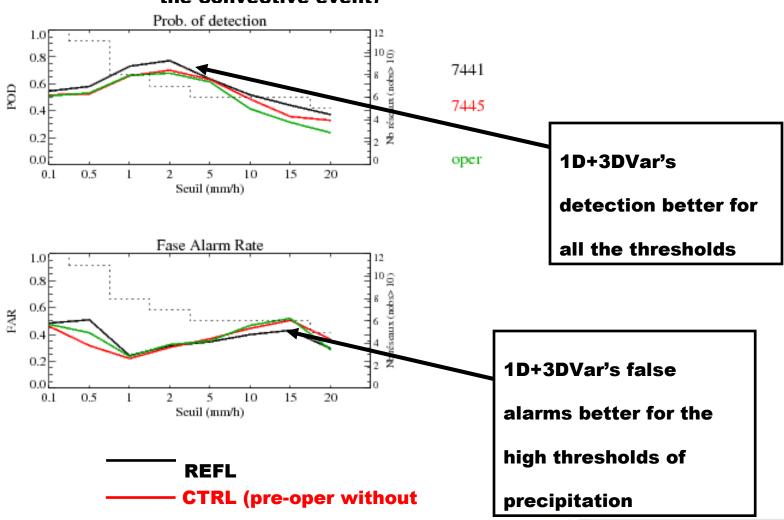




Average of QPF scores (for 6-hour forecast during

7/8 october



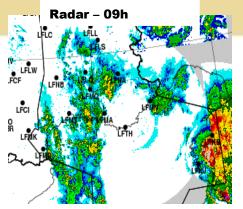


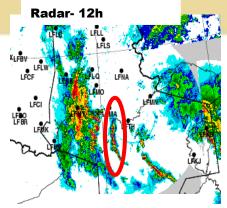
Dopples ré-OPER

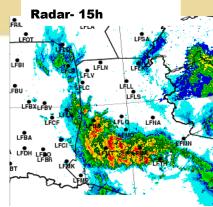


Case of the 04 no vember 2008: convection on mediterranean sea

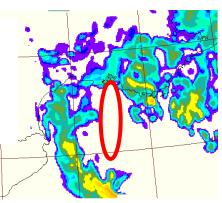
Chronology of the convective event







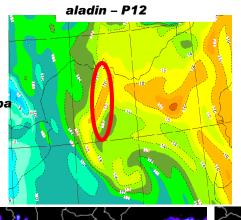
Simulated reflectivities from Arome, 12-hour forecast (850hpa)

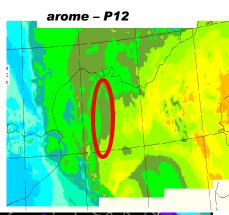


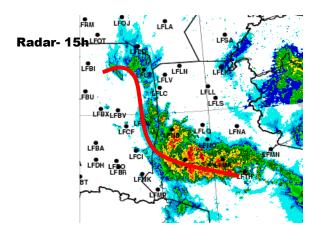
TPW – 850 hpa

(from medit.

South)





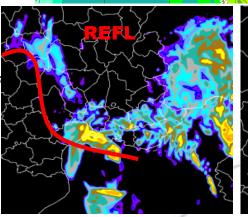


Simulated

Reflectivities from

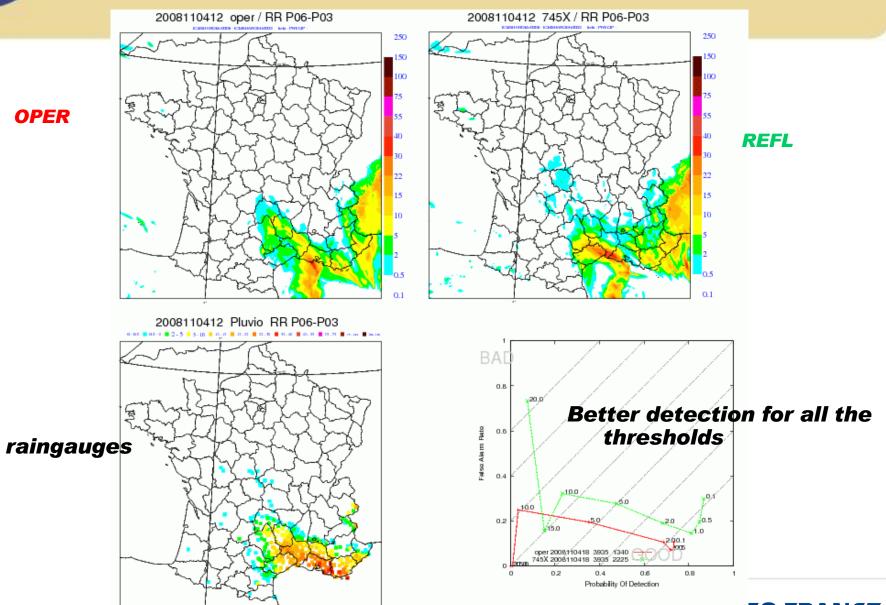
Arome, from 3
hour forecast

(900hpa) – r12

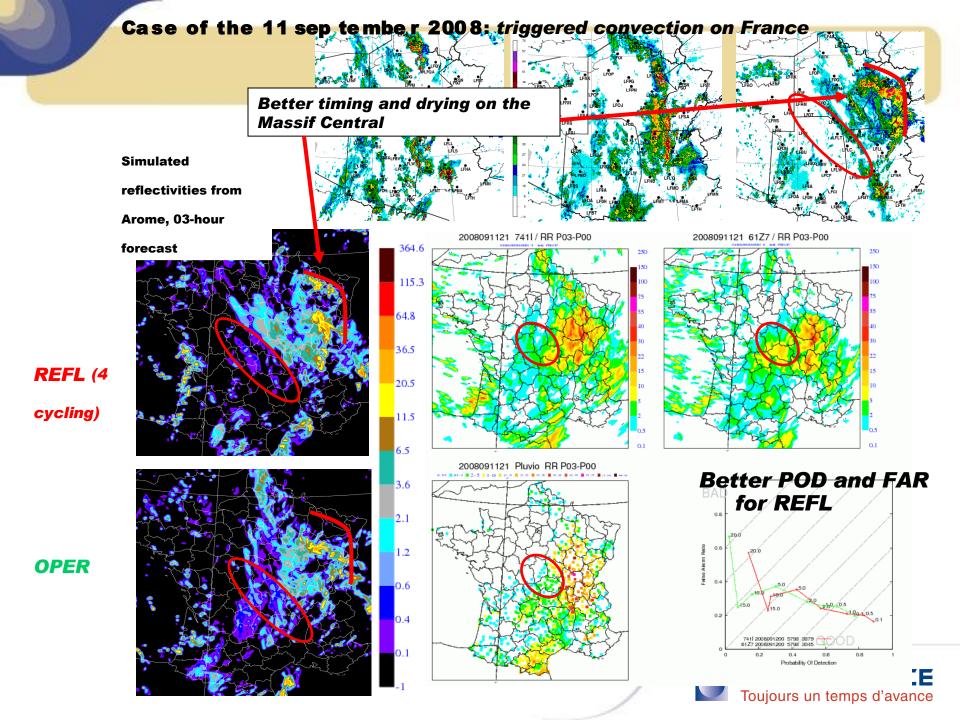




Case of the 04 no vember 2008: convection on mediterranean sea

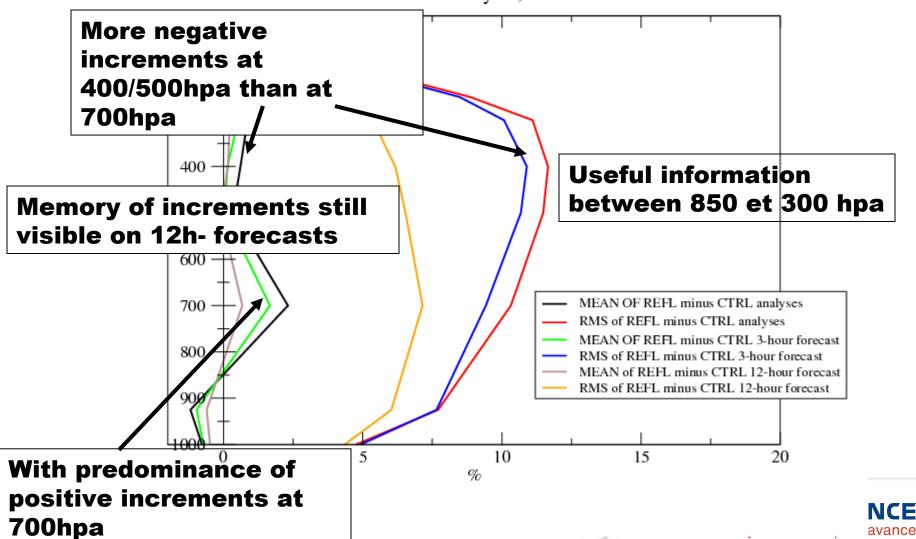




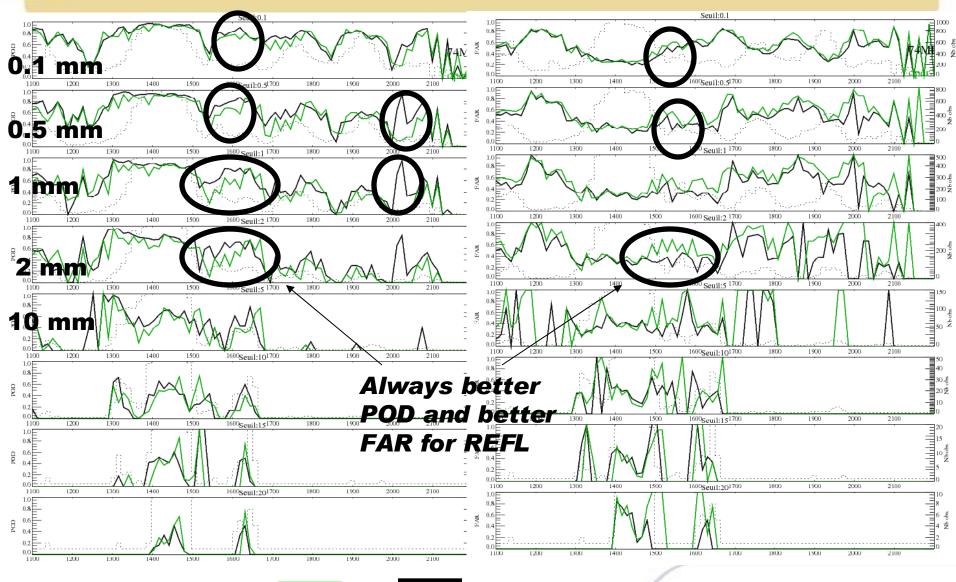


Analysis increments: behavior through the short-term forecasts (over 1,5 month)

One month average and rms of normalized relative humidity differences between REFL and CTRL from analyses, 3-hour and 12-hour forecast



QPF evaluation: Good detection/False Alarme...

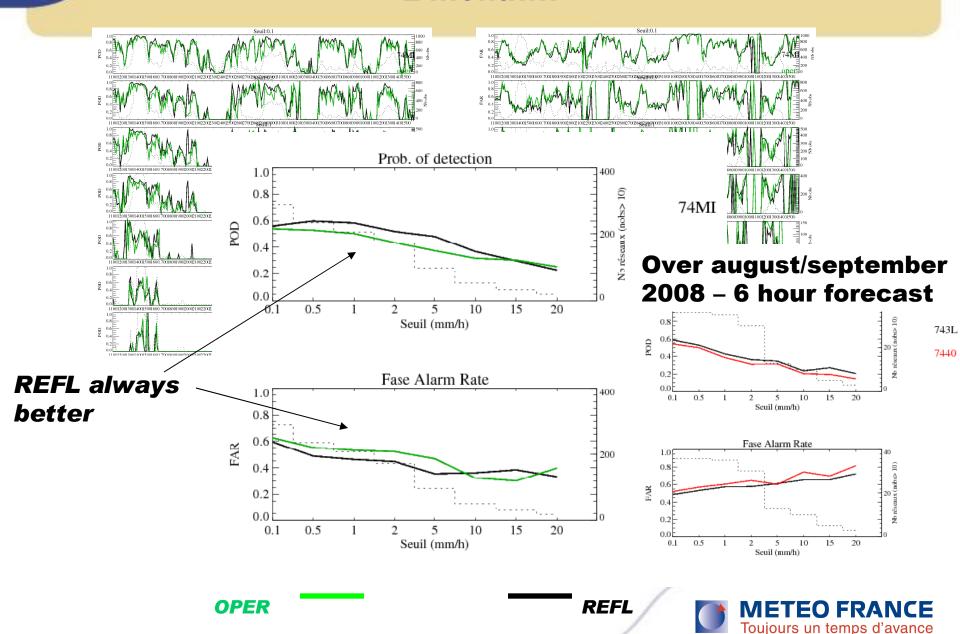


OPER

REFL



QPF evaluation: POD/FAR: average on 2 month...



Humidity: difficult to validate with conventional data...

1000

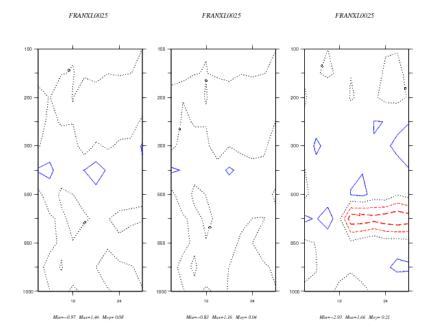
1.6

4.8

Scores with respect to own analysis, 12h-forecast

10 simulations (500 hPa) de 30 h du 20081211 au 20081221

EQM StD |biais|





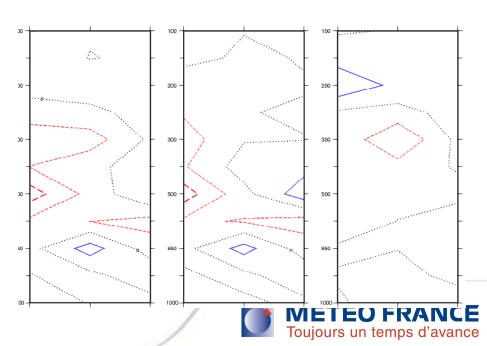
background departure o-b(ref)

background departure o-b

analysis departure o-a(ref)

x 0.0001

Scores with respect to RS, 12h-forecast



But good impact on wind field

VENT Echeance: 12 H Niveau: 500 hPa

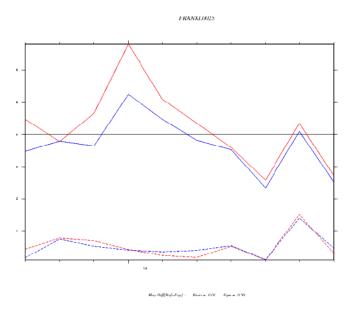
(m/s)

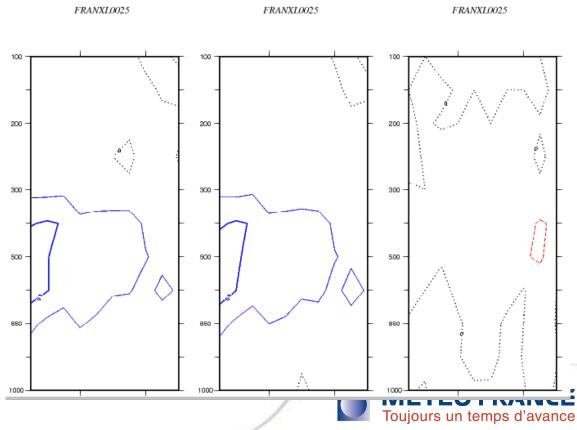
10 simulations de 21 h du 20081211 au 20081220

___Eqm P74AZ.r 00/A74AZ ___Eqm P74J2.r 00/A74J2
- - BiaisP74AZ.r 00/A74AZ -- BiaisP74J2.r 00/A74J2

With respect to RS, 24-forecast minus RS difference

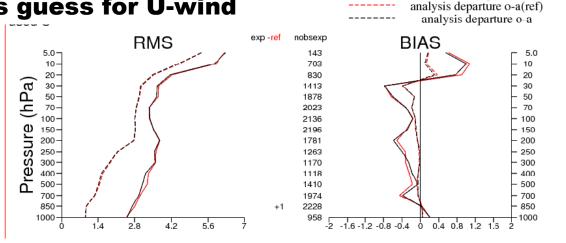
Eqm Ect Biais





Good impact on wind scores

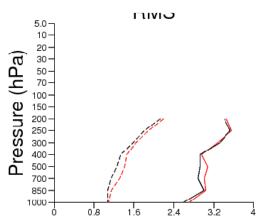
With respect to RS, improvement of the obs minus guess for U-wind



+1

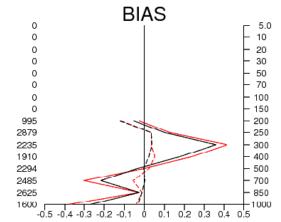
+3

With respect to Airep, improvement of the obs minus guess and analysis for Uwind



background departure o-b(ref)
background departure o-b
analysis departure o-a(ref)
analysis departure o-a

background departure o-b(ref) background departure o-b





Conclusion ☐ Sensitivities studies (thinning, observation error...) have shown robust results of the 1D+3DVar method ☐ But importance of a specific quality control and no-rain signal in particular for convective situations (check of the 1D convergence) ☐ Assimilation of reflectivities have been tested over two long periods (2) months last winter and 1.5 month last summer): systematic good QPF scores until 12hour forecast, reliable subjective scores on case studies, relatively neutral classical scores (positive impact on wind).

Perspectives of reflectivities assimilation

□ work is underway to incorporate reflectivities into the second operational version of AROME
☐ work to optimize assimilation of both radar winds and reflectivities together (not shown but less fit to radar winds with reflectivities)
☐ Test of the impact of the use of a flow-dependency B matrix or a clear sky/rainy B matrix



Bibliography

□ Caumont O, V. Ducrocq, G. Delrieu, M.Gosset, J-P. Pinty, J. Parent du Châtelet, H. Andrieu, Y. Lemaître, and G. Scialom, 2006: A radar simulator for high resolution Non-hydrostatic Models, journal of Atmospheric and Oceanic Technology, Vol. 23, 1049-1067.
□ Caumont O, E. Wattrelot, V. Ducrocq, G. Jaubert, F. Bouttier: First results of 1D+3DVar assimilation of radar reflectivities: <i>Proc. ERAD 2006,</i> Barcelona, 539
☐ Caumont, O., 2007 : Simulation et assimilation de données radar pour la prévision de la convection profonde à fine échelle. Thèse de l'Université de Toulouse, 252 p
☐ Montmerle, T_and C. Faccani, 2008: Mesoscale assimilation of radial velocities from Doppler radar in a pre-operational framework. <i>Mon. Wea. Rev, in press</i>
☐ Wattrelot E., O. Caumont, S. Pradier-Vabre, M. Jurasek and G. Haase, 2008: 1D+3Dvar assimilation of radar reflectivities in the pre-operational AROME model at Météo-France <i>Erad2008</i> , <i>Helsinki (Finland)</i>



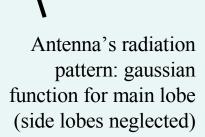
Reflectivity Observation operator • Bi-linear interpolation of the simulated (Caumont & al. 2006)

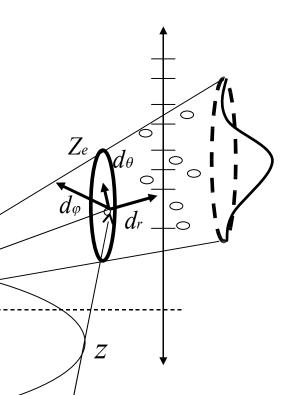
- hydrometeors (T,q, q_r, q_s, q_g)
- Compute radar reflectivity on each model level

 Simulated Reflectivity factor in « beam volum bv»

$$Z_e = 10\log(\int_{bv} \eta(r).f^4(\theta, \varphi).dr.d\theta.d\varphi)$$

Resolution volum, ray path: standard refraction (4/3 Earth's radius) and gate length is 250m, smaller than model resolution





Problematic of « no-rain » s ignal: is it sufficient to take into account the threshold of detection to well characterize the « no-rain » signal?

Problematic:

- if the SNR (signal-noise-ratio) is very low (bad quality of the radar or very far away from the radar), there is a high probability to dry with pixels below the noise but rainy!!
- But we know the noise, so we can ... take into account (in the 1D method) the threshold of detection for each pixel of each radar in order to not dry the model below this value: but is it sufficient?

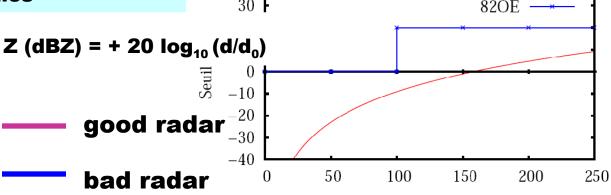
Use only $ZZ_{SIM} > ZZ_{THR}$ and (; Threshold of detection, function of range

40

30

Sensibilities studies have been done by deterioration

of one radar





Courbe Theorique

Range from the radar