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current DA method: nudging

radar-derived rain rates:

- latent heat nudging
- 1DVar (comparison) → will eventually be stopped

nudging radar radial wind V_r

PP KENDA for km-scale EPS:
LETKF

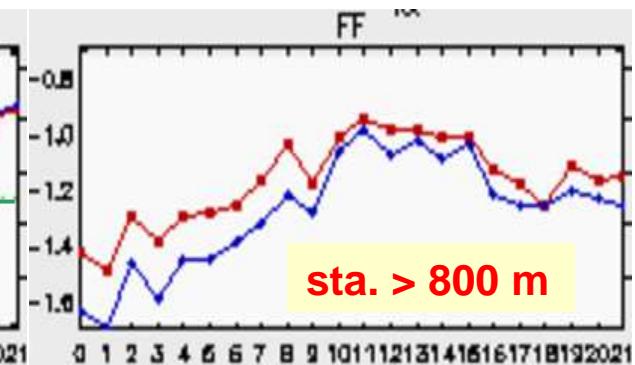
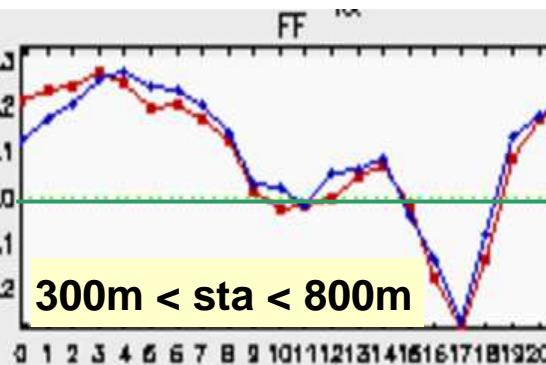
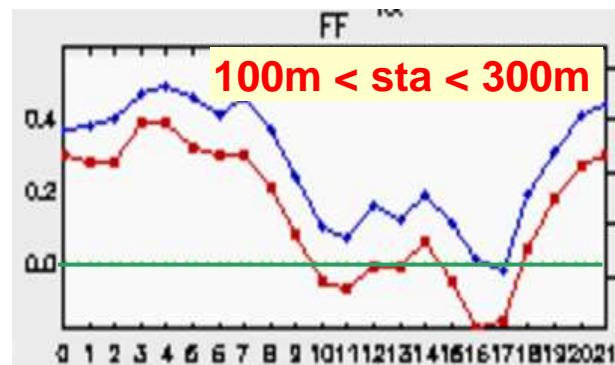
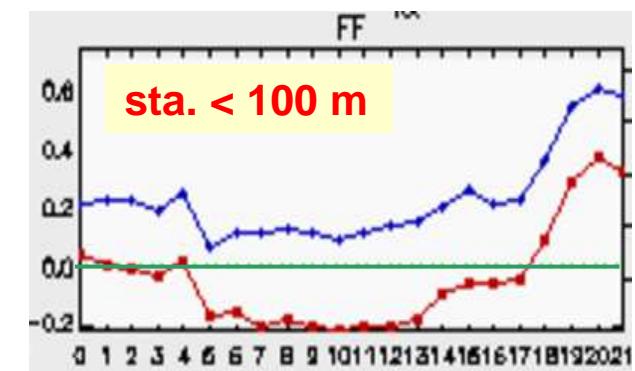
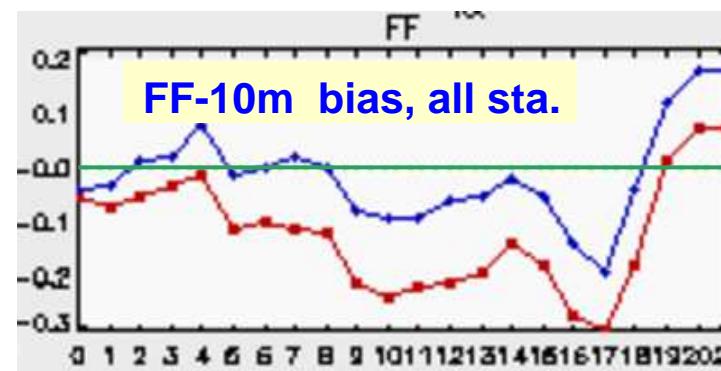
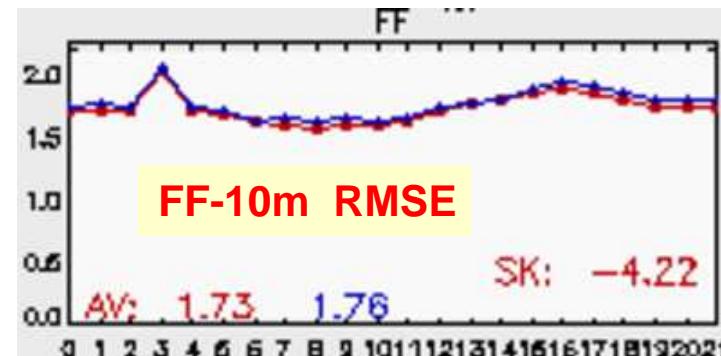
LETKF for HRM /
for 'large-scale' COSMO

nudging of radar radial velocity a side-step to verification



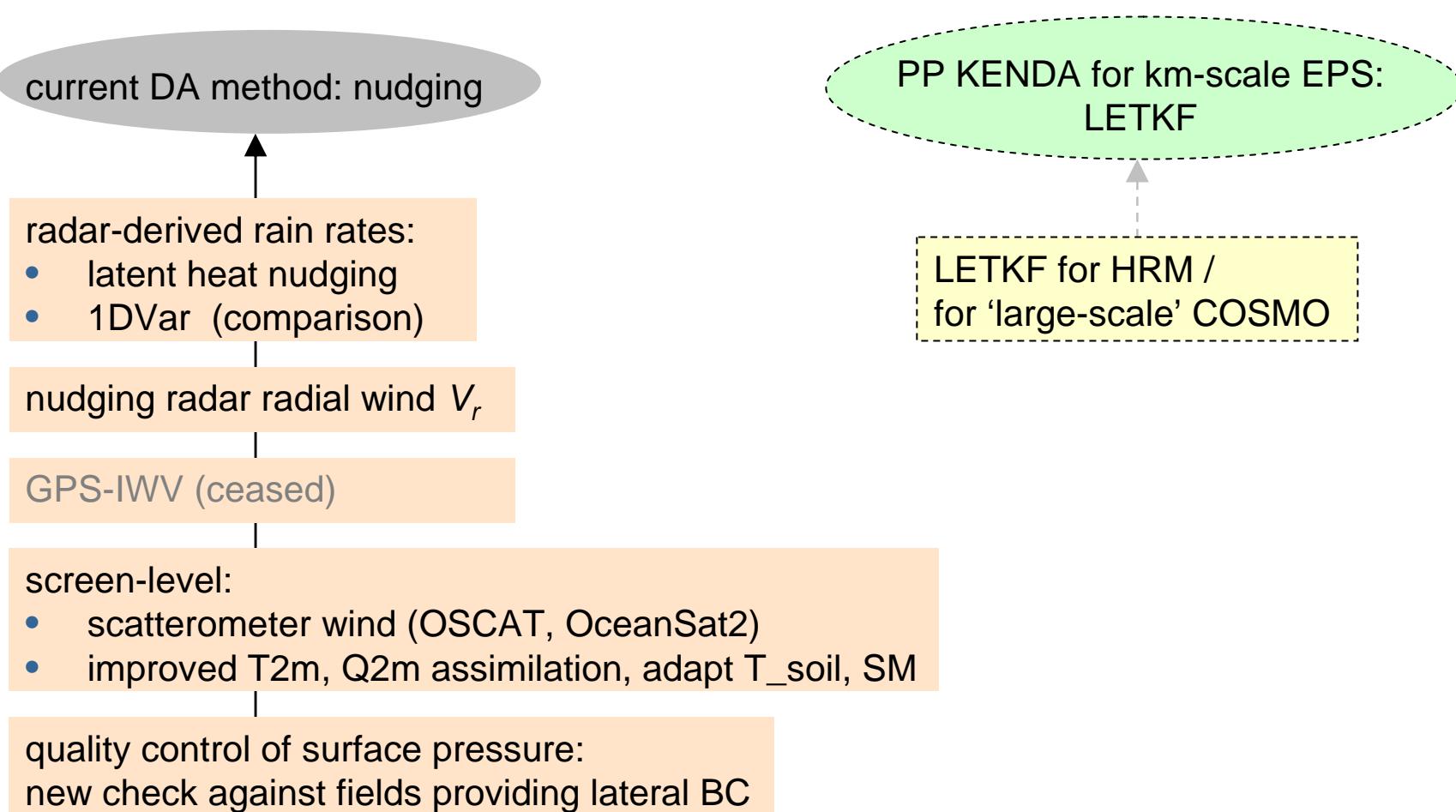
0-UTC runs

control
nudge vr



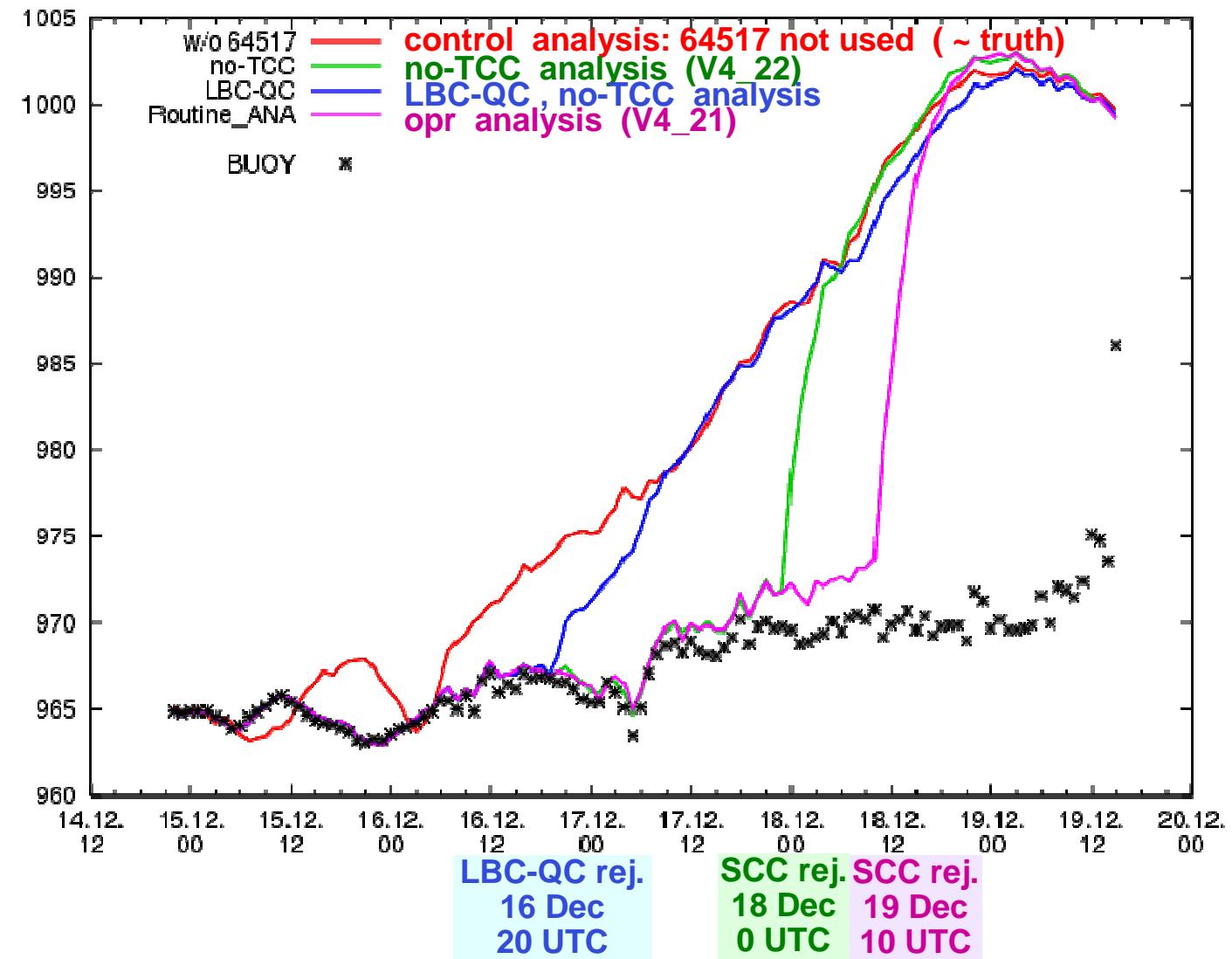
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quality control of surface pressure p_s :
case 2 : 15 - 19 Dec. 2011

surface
pressure
(at the location
of buoy 64517)



Status Overview on PP KENDA

Km-scale ENsemble-based Data Assimilation



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Contributions / input by:

Hendrik Reich, Andreas Rhodin, Roland Potthast, Uli Blahak, Klaus Stephan, Africa Perianez,
Michael Bender (DWD)

Annika Schomburg (DWD / Eumetar)

Yuefei Zeng, Dorit Epperlein (KIT Karlsruhe)

Daniel Leuenberger (MeteoSwiss)

Mikhail Tsyrulnikov, Vadim Gorin, Igor Mamay (HMC)

Lucio Torrisi (CNMCA)

Amalia Iriza (NMA)

- Task 1: General issues in the convective scale (e.g. non-Gaussianity)
- Task 2: Technical implementation of an ensemble DA framework / LETKF
- Task 3: Tackling major scientific issues, tuning, comparison with nudging
- Task 4: Inclusion of additional observations in LETKF

- modifications in COSMO in official code (V4_24)
(e.g. in order to have a sub-hourly update frequency)
- **COSMO-DE LETKF implemented in NUMEX** and tested
(e.g. stand-alone 2-day experiment reproduced)
- **GME-LETKF & ensemble INT2LM for DA cycle implemented in NUMEX**,
being tested, should be available end of Sept.
 - in Oct., start first KENDA experiments in NUMEX over several days/weeks
 - but:
 - direct interpolation from 60 km to 2.8 km !
 - deterministic analysis not yet implemented in NUMEX
- ensemble LBC 2013 – 2014 :
ensemble perturbations of interpolated ensemble GME fields,
added to deterministic COSMO-DE LBC

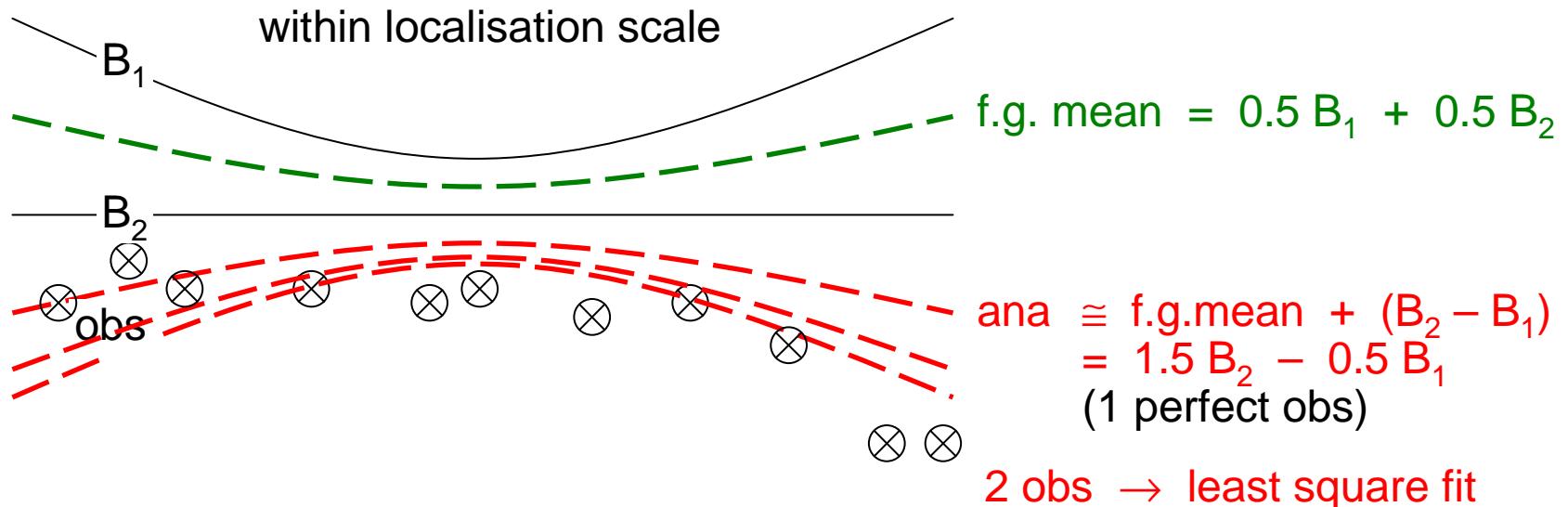
- in past year, still only preliminary LETKF experiments possible, using Hendrik's scripts:
 - up to 2 days (7 – 8 Aug. 2009: quiet + convective day)
→ 3-hourly (15-min) cycles
 - 32 ensemble members
 - perturbed LBC: COSMO-SREPS, 3 * 4 members

- therefore
 - theoretical studies, toy model experiments related to adaptive localisation
→ talk by Hendrik Reich
 - benchmark , winter school on DA , support for HERZ centre , testing (e.g. NUMEX) ...
 - only few COSMO-DE experiments related to adaptive localisation

- **tool for production of ‘full’ NetCDF feedback files**
 - make clean interfaces to observation operators / QC in COSMO : done
 - ... integrate them into 3DVAR package : in progress
 - and extend flow control (read correct (hourly) Grib files etc.) : to be done

should be ready by end of 2012 (for VERSUS)
- **ensemble-related diagnostic + verification tool**, using feedback files :
(Iriza, NMA)
 - computes statistical scores for different runs ('experiments'),
 - **focus: use exactly the same observation set in each experiment !**
 - select obs according to namelist values (area, quality + status of obs, ...)
 - implementing ensemble scores (reliability, ROC, Brier Skill Score, (continuous) Ranked Probability Score)
 - main part of documentation written

- adaptive estimation of obs error covariance \mathbf{R}
(Li, Kalnay, Miyoshi, QJRMS 2009) , but our implementation: in ensemble space !



add obs, if already $N_{obs} > N_{ens}$:

- cannot be fitted well, improve analysis only slightly

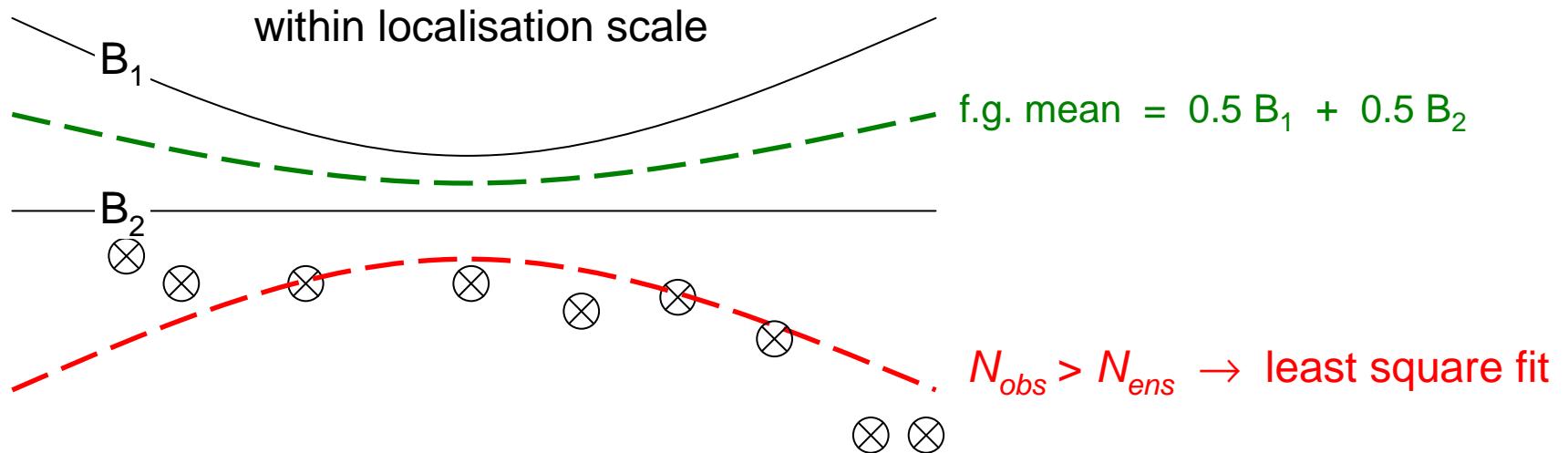
- decrease analysis error !

$$\mathbf{P}_w^a = \left[(k-1) \mathbf{I} + (\mathbf{Y}^b)^T \mathbf{R}^{-1} \mathbf{Y}^b \right]^{-1}$$

→ adaptive \mathbf{R} takes that into account and increases \mathbf{R}



however: large N_{obs} : adaptive increase of \mathbf{R} indicates non-optimal use of obs



→ localisation !

→ see also Hendrik's talk !

(or data selection / superrobbing ?)

→ basic idea for adaptive localisation : keep N_{obs} constant ($> N_{ens}$, not $>> N_{ens}$) !

LETKF, preliminary results: horizontal localisation

Caspari-Cohn function: scale $s = 100 \text{ km}$

$$\rightarrow 0.4 \text{ at } r \approx (2)^{1/2} \cdot s \approx 141 \text{ km}$$

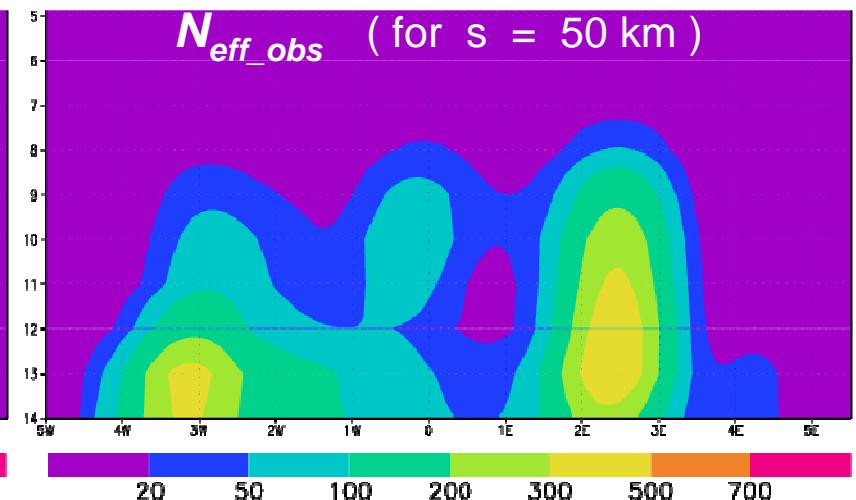
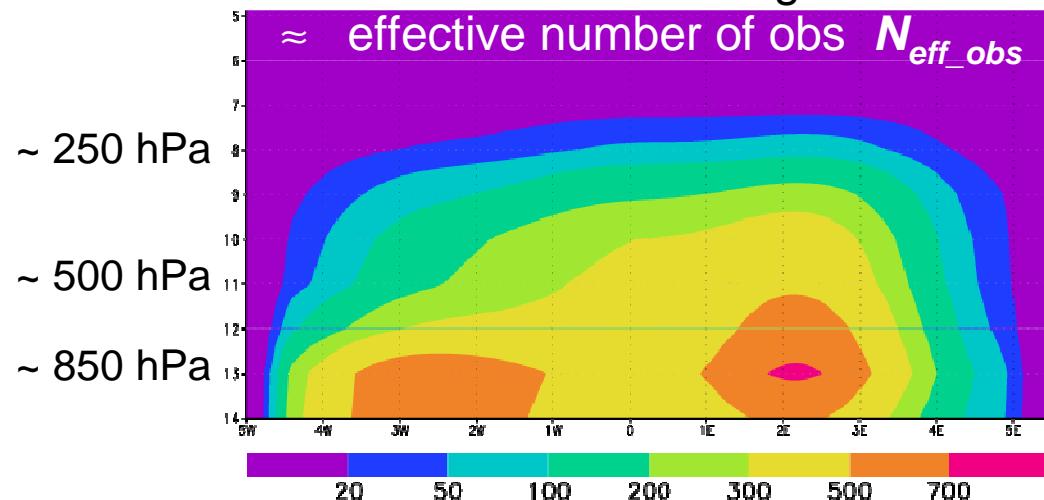
$$\rightarrow 0 \text{ at } r = 2 \cdot (10/3)^{1/2} \cdot s \approx 365 \text{ km}$$

→ reduce scale : $s = 50 \text{ km}$

vertical cross section

(at rot lat = 2° , 8 Aug 2009, 12 UTC)

sum of localisation weights of obs

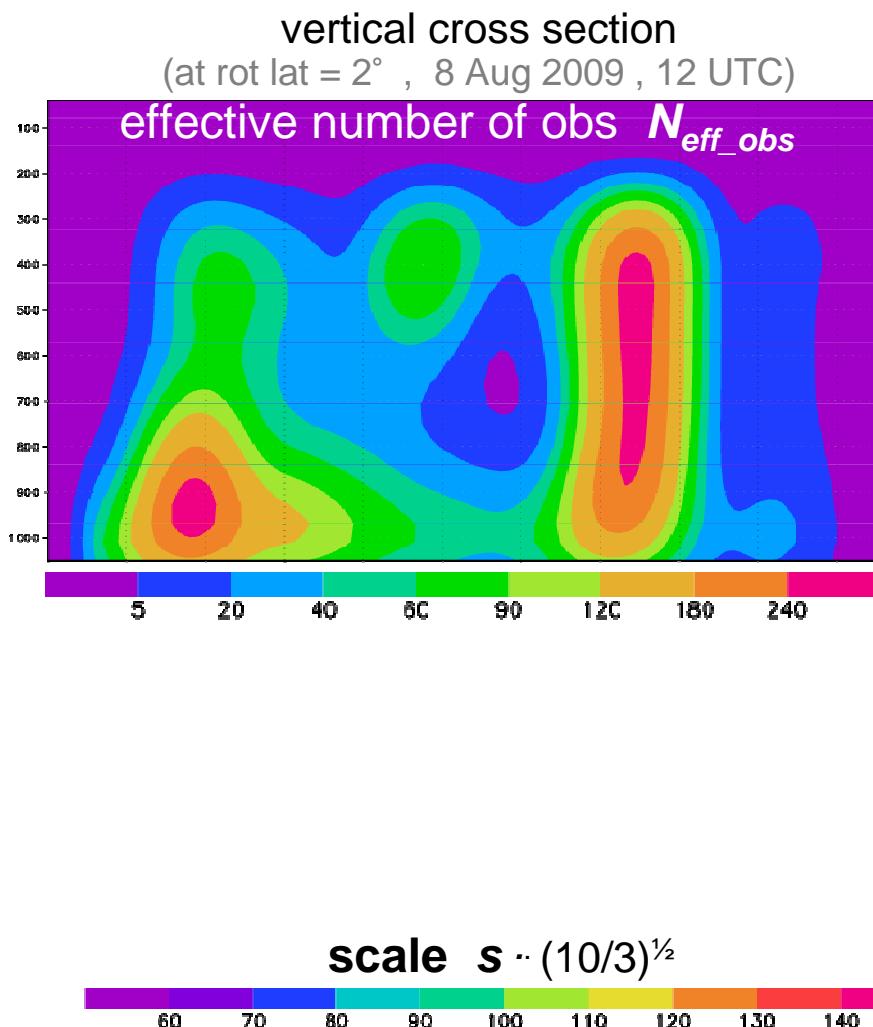


→ $N_{\text{eff_obs}} \gg N_{\text{ens}}$

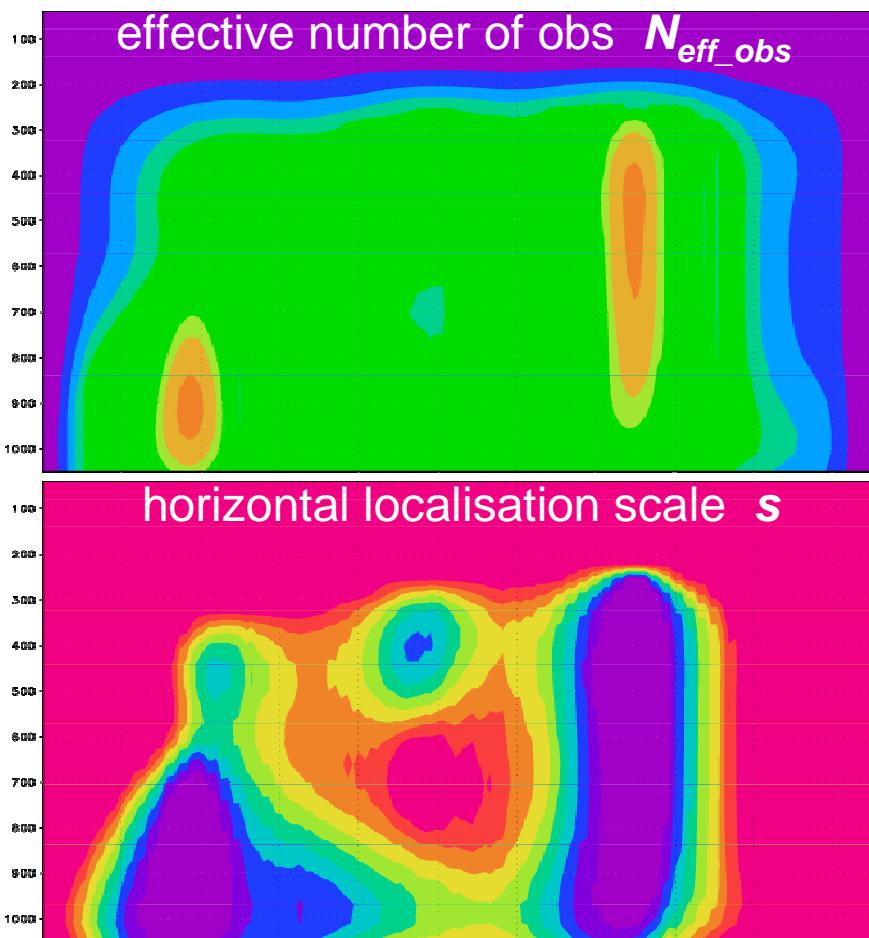
→ too few degrees of freedom
in order to fit the observations

LETKF, preliminary results: adaptive horizontal localisation

Caspari-Cohn function: scale $s = 50 \text{ km}$



→ adaptive scale s :
adapt s such that $N_{\text{eff_obs}} \approx 70$
and $30 \text{ km} \leq s \leq 80 \text{ km}$



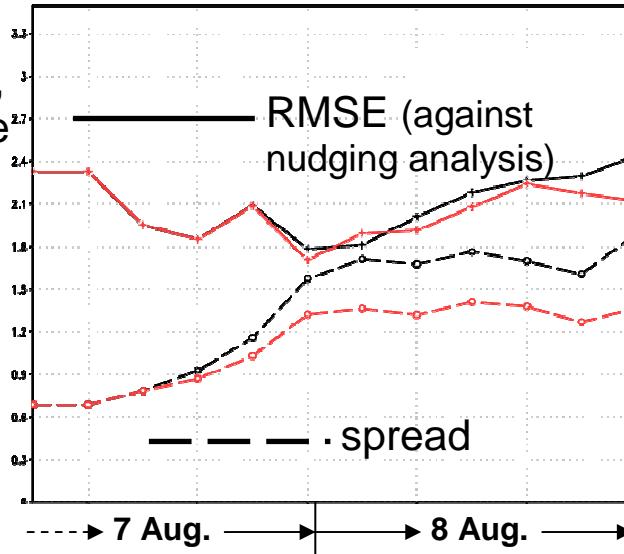
LETKF, preliminary results: adaptive horizontal localisation

first guess mean
(inner domain average)
(variable vertical localisation,
adaptive R and multiplicative
covariance inflation ρ)

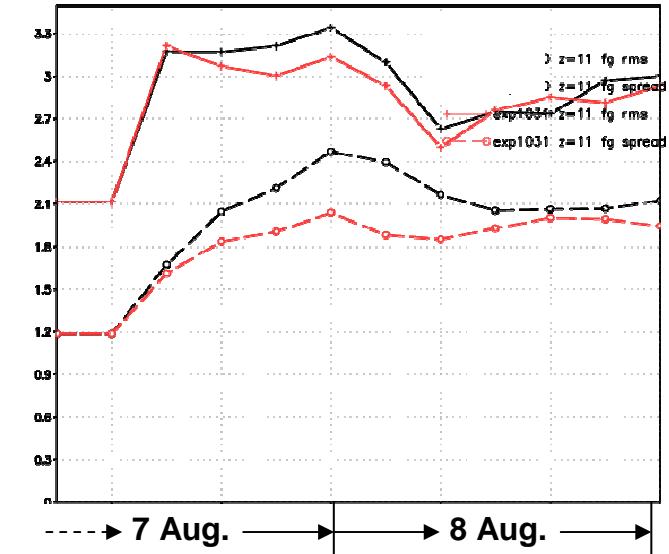
$s = 50 \text{ km}$
adaptive s

→ smaller spread
→ mostly smaller RMSE
(mixed results in verif
vs. upper-air obs
(T pos., wind neg.))

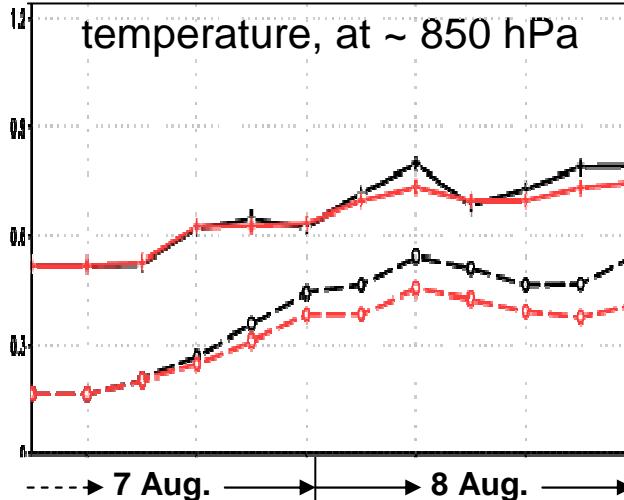
zonal wind, at $\sim 850 \text{ hPa}$



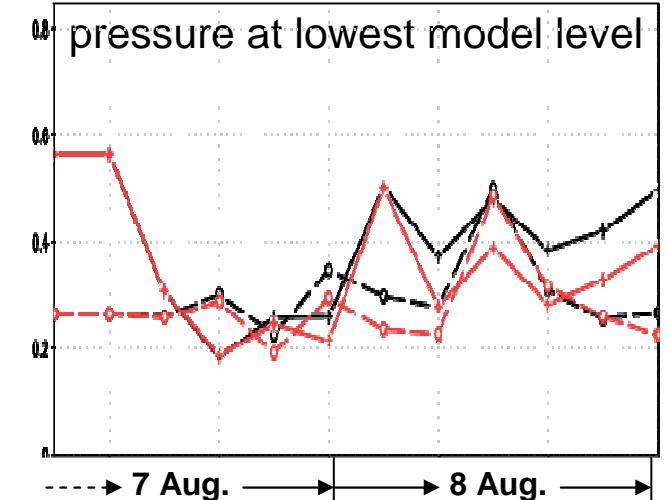
zonal wind, at $\sim 250 \text{ hPa}$



temperature, at $\sim 850 \text{ hPa}$



pressure at lowest model level



- **parameterisation of model error using statistics** (Tsyrulnikov, Gorin) :

- parameterisation: $e = \mu * F_{phys}(x) + e_{add}$
- estimate parameters by fitting to statistics from forecast **tendency** and observation **tendency** data (using a maximum likelihood based method)

failed in OSSE setup with simulated ME for finite-time 1 – 6 hr tendencies !!!

main methodological cause of failure : instantaneous ME is contaminated
in **finite-time** tendencies
by other tendency errors :

- trajectory drift as a result of ME themselves
 - initial errors (plus the trajectory drift due to initial errors)
- conclusion: observation accuracy and spatio-temporal coverage
far from being sufficient to reliably estimate ME !
- task is stopped !

→ new task for a **pattern generator** (PG)

purely stochastic tool to generate 4-D pseudo-random fields with selectable scales / ampl.,
used to generate additive perturbations / for stochastic physics
(~ 0.4 FTE / y)

- **stochastic physics:** perturbing total physics tendency by a random factor
at any given grid point (Palmer et al., 2009) (Torrisi)
 - modified Buizza version running
 - tuning required
 - perturb all physics tendencies in same way ?
- 2013 Ekaterina Machulskaya from SFP for (more physically based) stochastic physics !
+ 1 N.N. (renewable energy project)
- additional additive inflation: - by scaled forecast differences (e.g. Bonavita et al.) ?
- 3DVAR – B ?

- **radar : assimilate 3-D radial velocity and 3-D reflectivity directly**
 1. observation operators implemented
 - (Uli Blahak (DWD), Yuefei Zeng, Dorit Epperlein (PhD, KIT))
 - full, sophisticated
 - efficient (e.g. lookup tables for Mie scattering)
 - tested for sufficiently accurate and efficient approximations for DA
 2. assimilation experiments
 - technical work finished this week
 - 1 - 2 assimilation case studies (Zeng)
 - 2013: Klaus Stephan : test periods, tuning ...



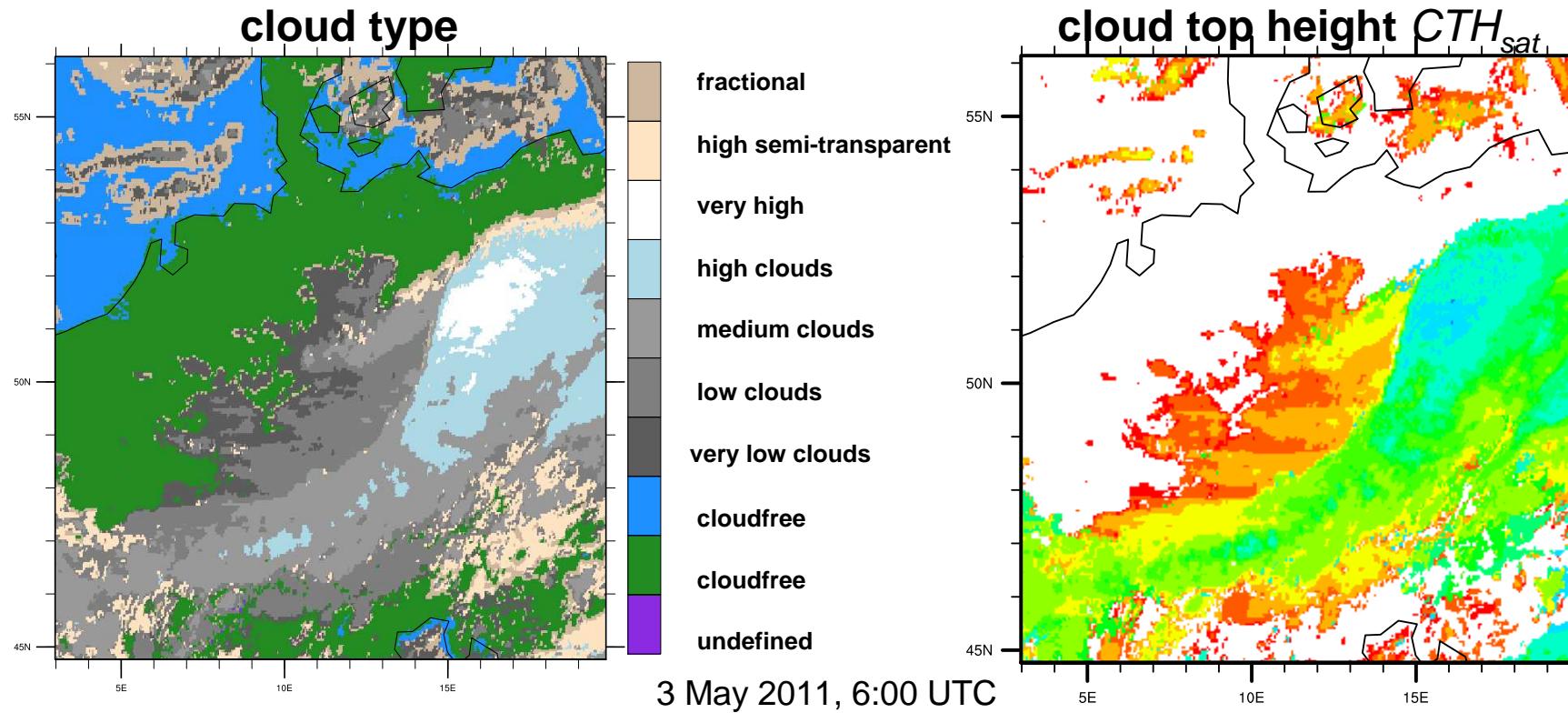
- **ground-based GNSS slant path delay SPD**
 - produce & use **tomographic** refractivity profiles (Erdem Altunac, PhD)
 - implement non-local **SPD** obs operator & use SPD (Michael Bender)
 - first implement SPD obs operator in 3DVAR package
(environment for work on tomography)
 - implement simple operator (refractivity along straight line)
 - implement complex obs operator with ray tracer
 - monitoring, test e.g. impact of straight line approximation
 - then implement obs operator in COSMO (in 2013)



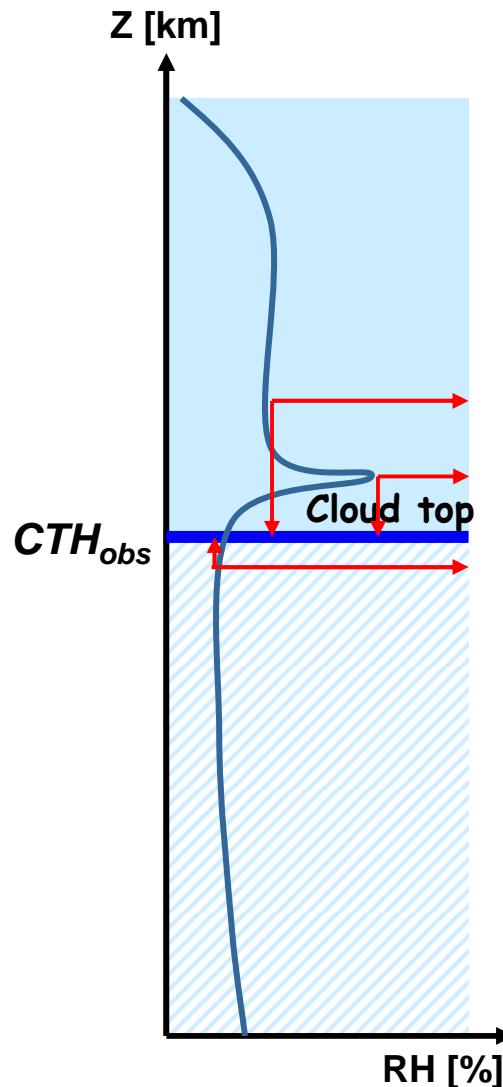
- **cloud information based on satellite and conventional data**
 1. derive incomplete analysis of cloud top + base height, using conventional obs (synop, radiosonde, ceilometer) and NWC-SAF cloud products from SEVIRI
use cloud top height info in LETKF
(Annika Schomburg , DWD / Eumetsat)
 2. use SEVIRI **brightness temperature** directly in LETKF in cloudy (+ cloud-free) conditions, in view of improving the horizontal distribution of cloud and the height of its top (2013: Africa Perianez, Annika Schomburg)

→ compare approaches

Particular issues: non-linear observation operators,
 non-Gaussian distribution of observation increments



- retrieval algorithm needs temperature and humidity profile from a **NWP model**
 → **cloud top height CTH_{sat} wrong if temperature profile in NWP model wrong !**
 → combine good horizontal resolution of satellite info
 with good vertical resolution of radiosonde info :
 use nearby radiosondes with same cloud type to correct CTH_{sat}



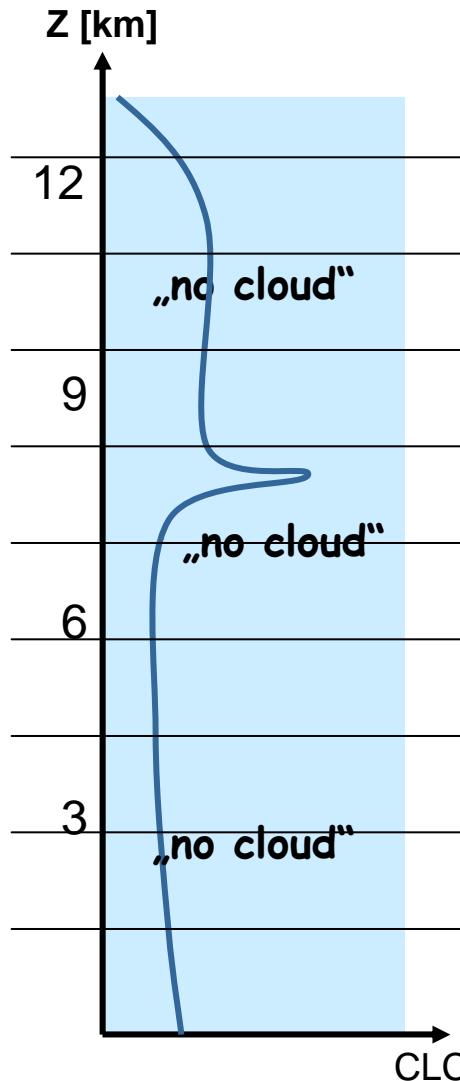
if cloud observed with cloud top height CTH_{obs} ,
what is the appropriate type of obs increment ?

- avoid too strong penalizing of members with high humidity but no cloud
 - avoid strong penalizing of members which are dry at CTH_{obs} but have a cloud or **even only high humidity** close to CTH_{obs}
- search in a vertical range Δh_{max} around CTH_{obs} for a 'best fitting' model level k , i.e. with minimum 'distance' d :

$$d = \min_k \sqrt{(f(RH_k) - f(RH_{obs}))^2 + \frac{1}{\Delta h_{max}}(h_k - CTH_{obs})^2}$$

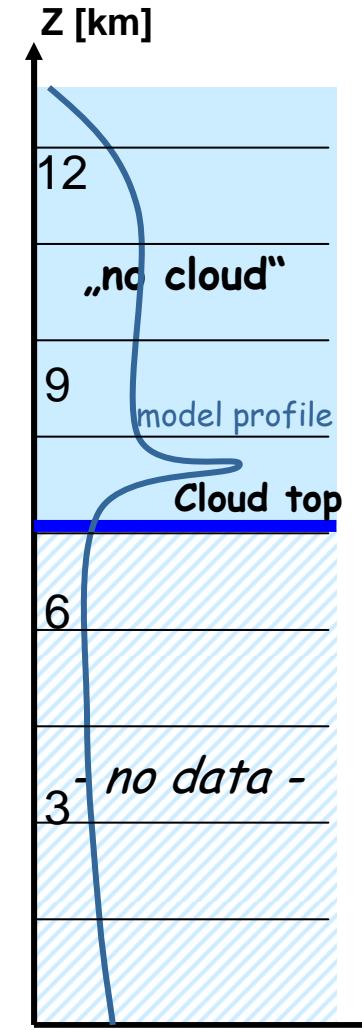
function of relative humidity
= 1
height of model level k

- use $f(RH_{obs}) - f(RH_k) = 1 - f(RH_k)$
and $CTH_{obs} - h_k$
as 2 separate obs increments in LETKF



type of obs increment ,
if **no cloud** observed ?

- assimilate $CLC = 0$ separately
for high, medium, low clouds
- model equivalent:
maximum CLC within vertical range

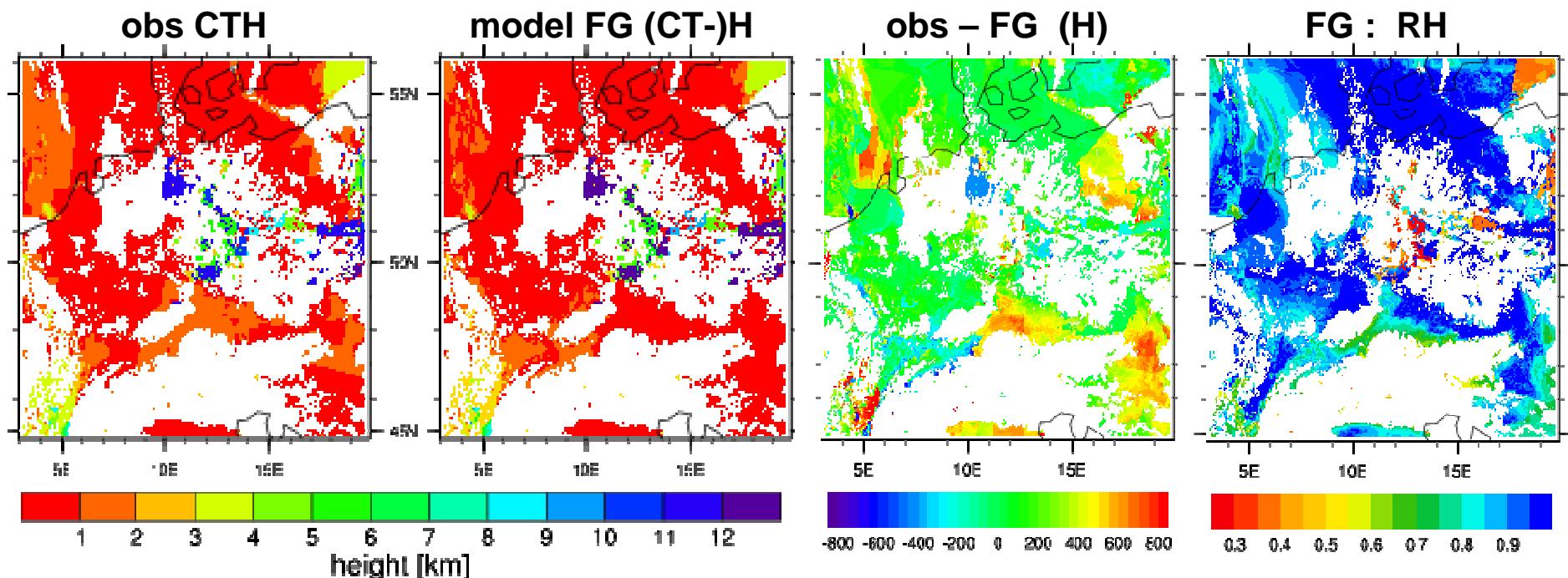


use of cloud info :
assimilation of 'cloud analysis' : example

Deutscher Wetterdienst
Wetter und Klima aus einer Hand

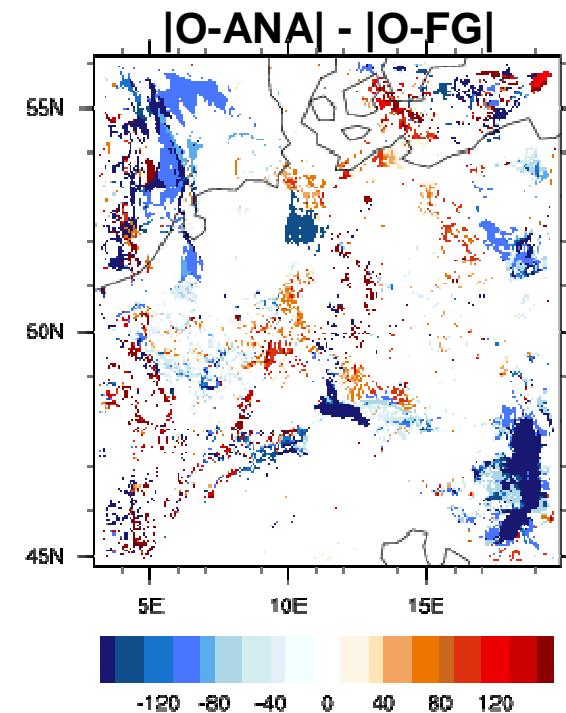
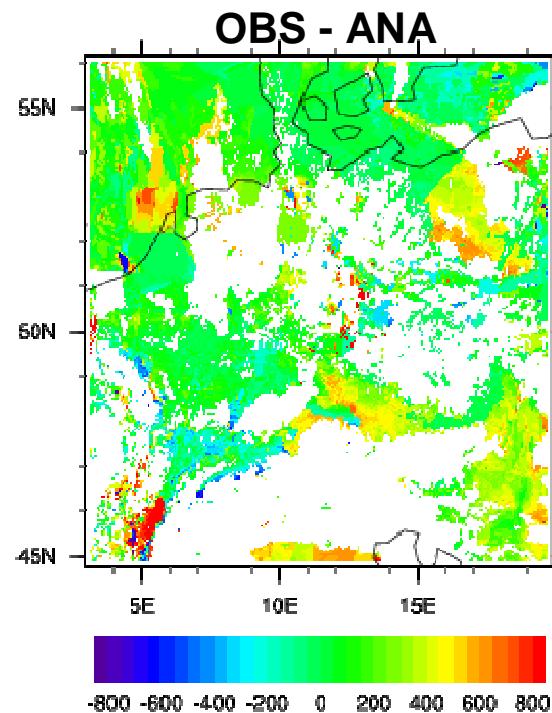
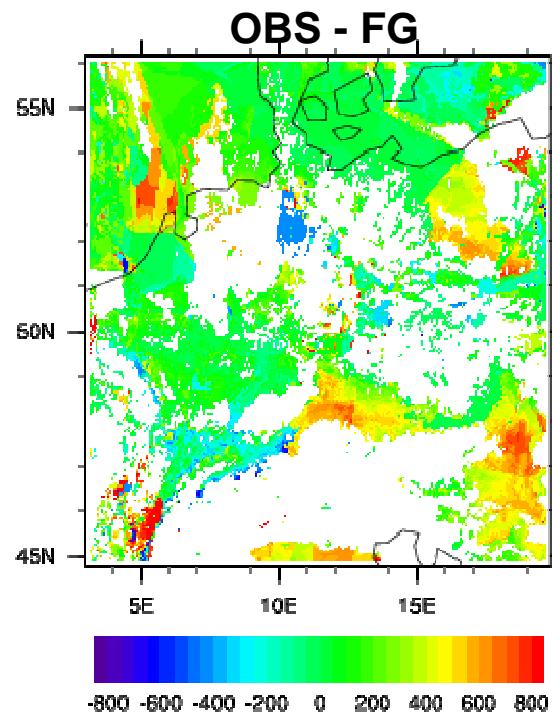


17 Nov 2011, 6 UTC (low stratus case)
pixels where observation has clouds
(output from feedback files)





'cloud' top height

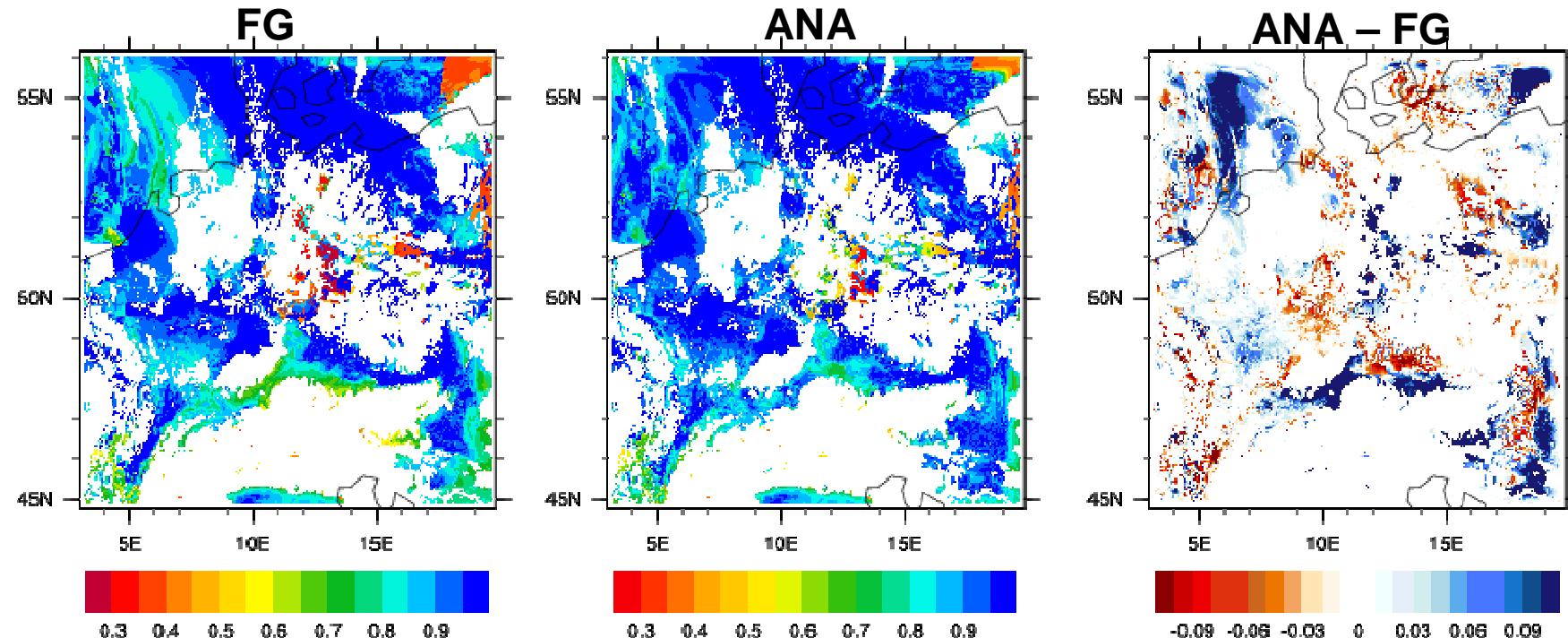


blue means ANA is better

Here: results of **deterministic** run in LETKF framework
(Kalman gain matrix applied to standard (unperturbed) model integration)

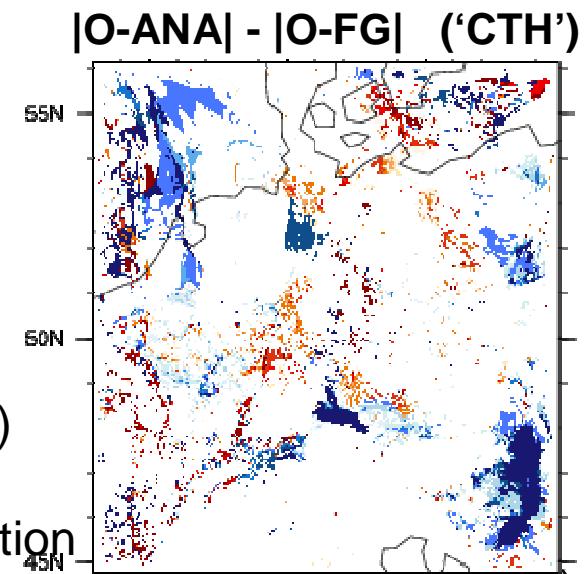
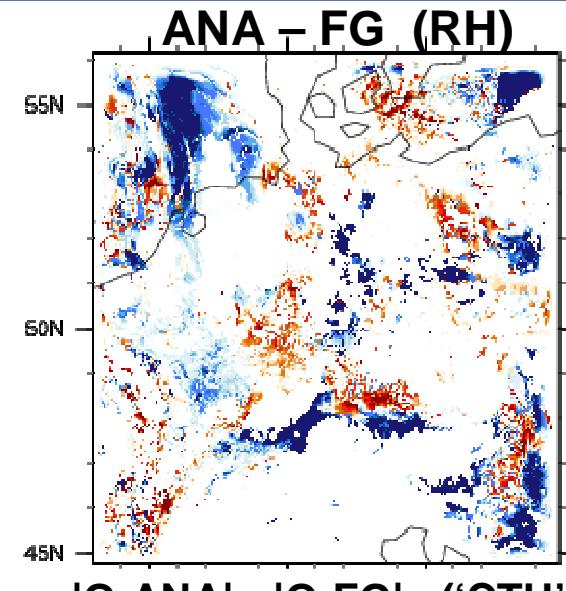
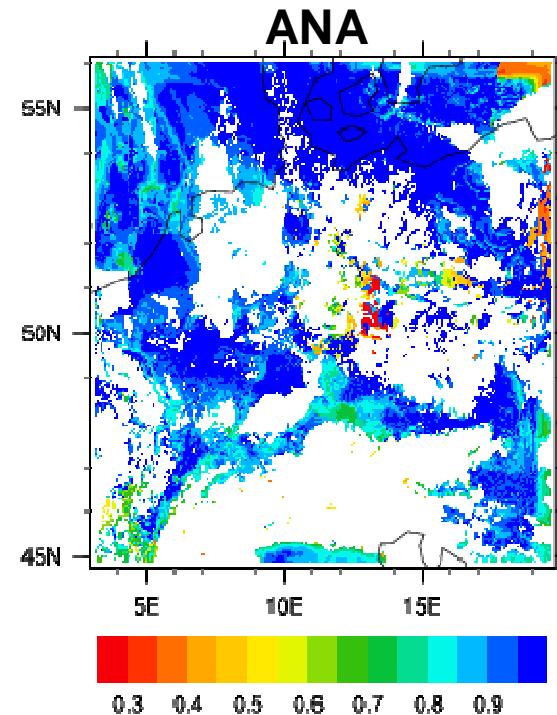
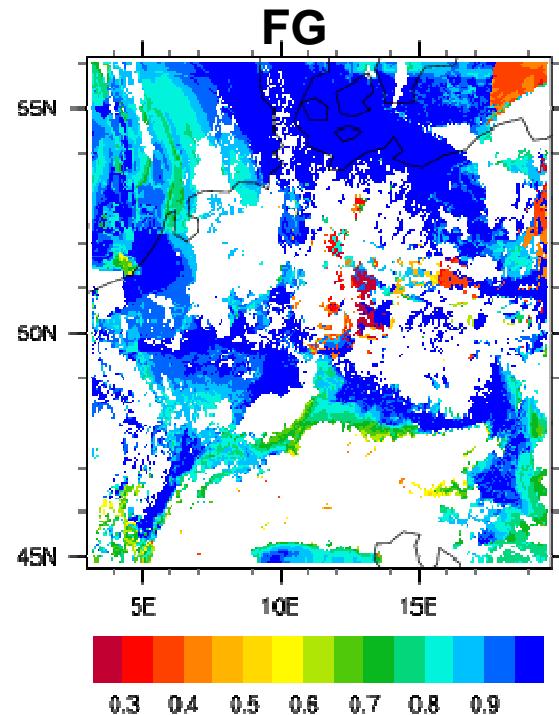


Relative humidity at 'cloud' level



Here: results of **deterministic** run in LETKF framework
(Kalman gain matrix applied to standard (unperturbed) model integration)

Relative humidity at cloud level



→ LETKF draws model cloud tops closer to obs

next : – detailed evaluation (cross section, profiles...)
– single observation experiments
– tuning of observation error, thinning, localization



thank you for your attention

