

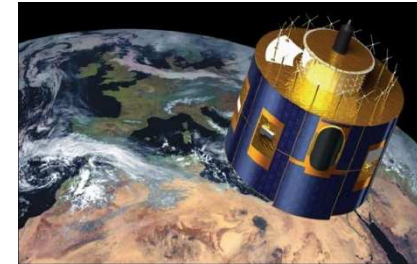


Assimilating cloud information into the COSMO model using the LETKF: Results from cycling and forecast experiments

Annika Schomburg, Christoph Schraff

This work was funded by the EUMETSAT fellowship programme.

- Geostationary satellite data: **Meteosat-SEVIRI**
($\Delta x \sim 5 \text{ km}$ over central Europe, $\Delta t = 15 \text{ min}$)



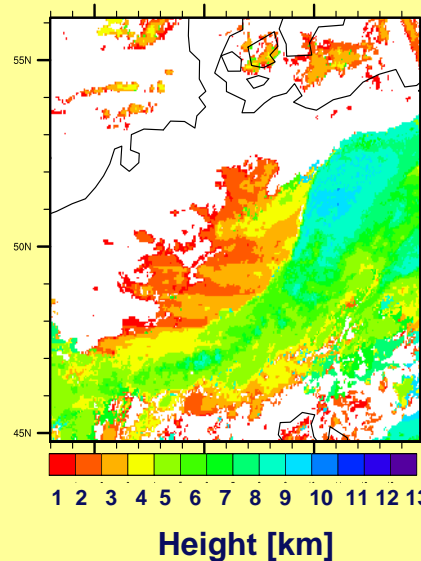
Source: EUMETSAT

NWCSAF Satellite product: cloud top height

retrieval algorithm
uses T, q_v profile
from NWP model as input

→ *cloud top height error
if T model profile
not correct*

→ *use also radiosonde
info where available*

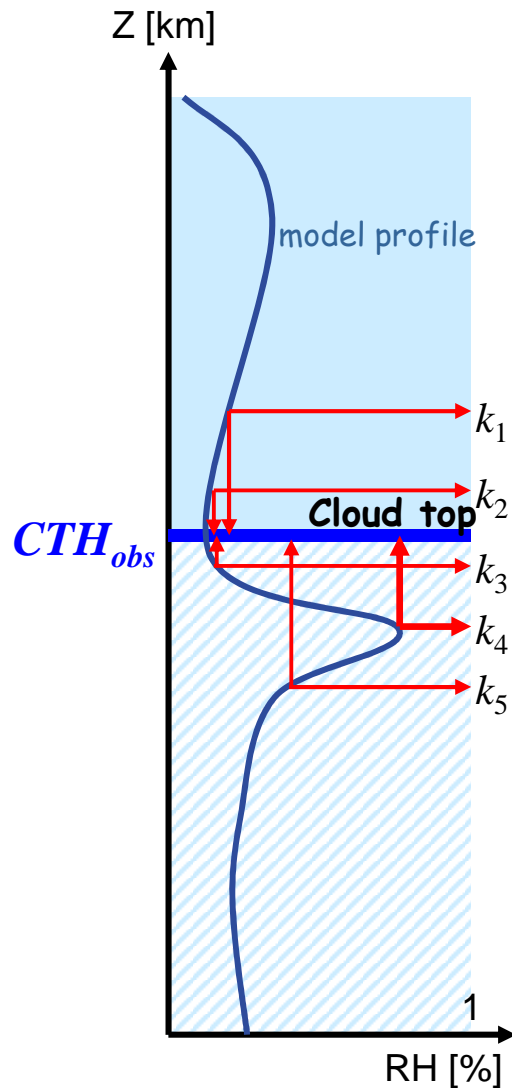


cloud top
height

relative
humidity
at cloud
top height

cloud
fraction

Use of cloud top height (CTH) 'obs' in LETKF: Method for **cloud-covered** pixels



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

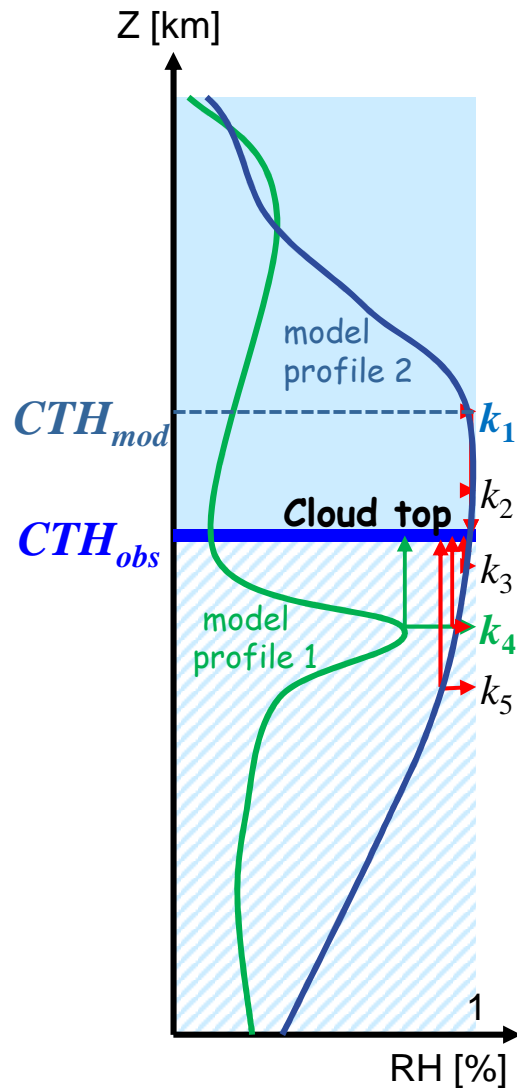
- avoid too strong penalizing of members with high humidity but no cloud (→ use RH instead of cloud fraction)
 - avoid strong penalizing of members which are dry at CTH_{obs} but have a cloud or **even only high humidity** close to CTH_{obs}
- search in a vertical range Δh_{max} around CTH_{obs} for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

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

↑ relative humidity ↑ = 1 ↑ height of model level k



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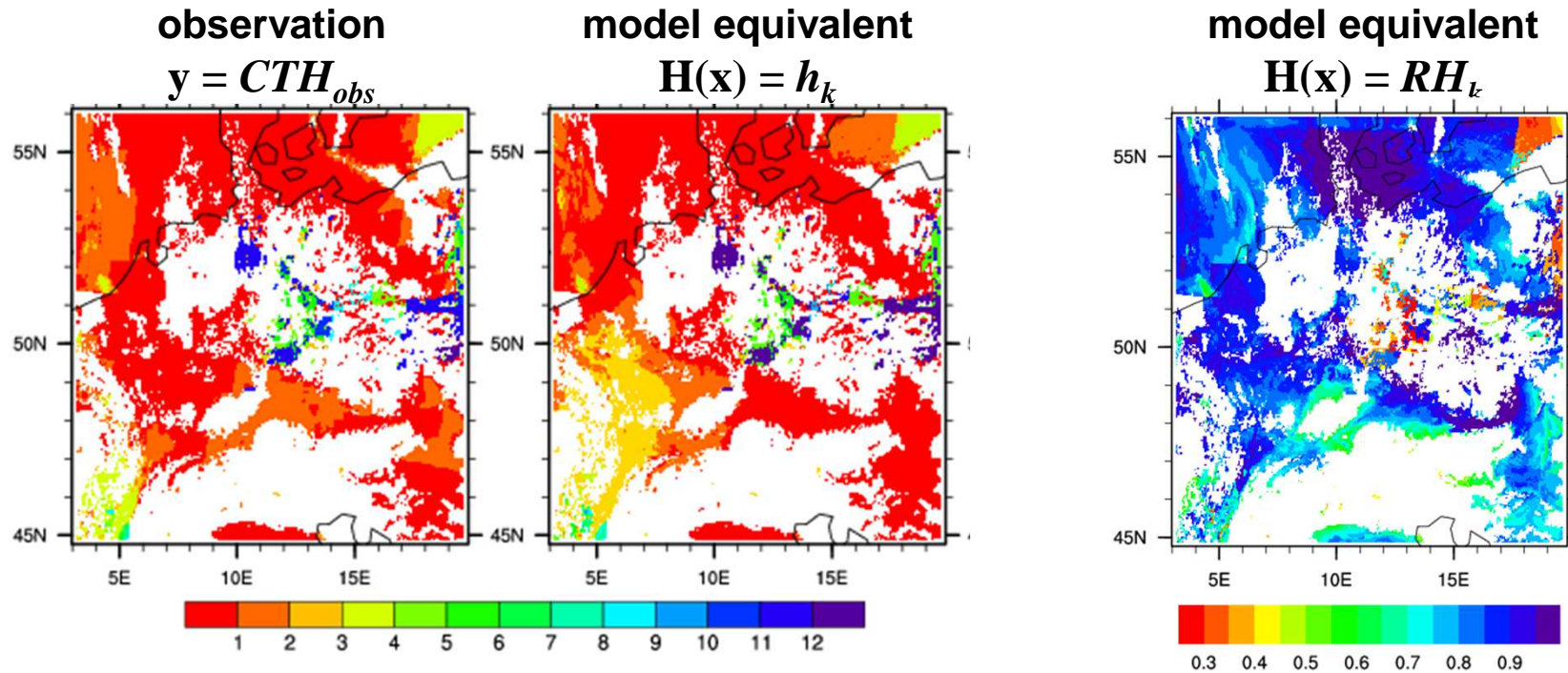
(but make sure to choose the **top** of the detected cloud)

- use $y = CTH_{obs}$, $H(x) = h_k$
and $y = RH_{obs}=1$, $H(x) = RH_k$ (over water/ice dep. on T)
as 2 separate variables assimilated by LETKF

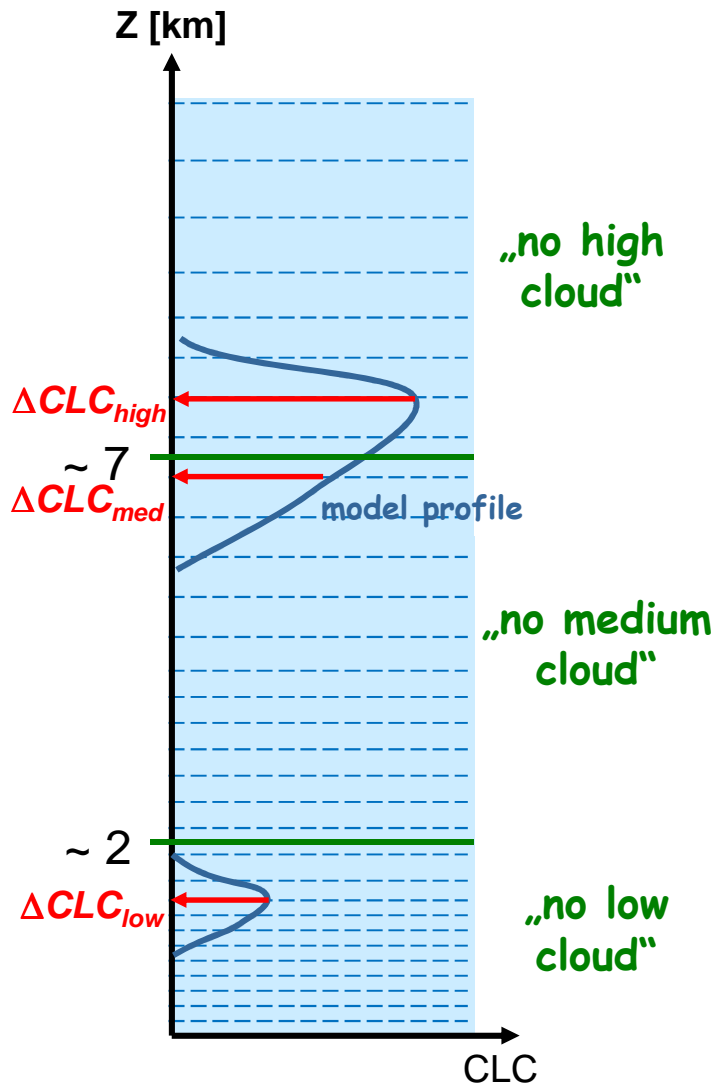


Use of CTH 'obs' for cloud-covered pixels : Example for obs / model equivalents

„Cloud top height“



Use of cloud top height (CTH) 'obs' in LETKF: Method for **cloud-free** pixels

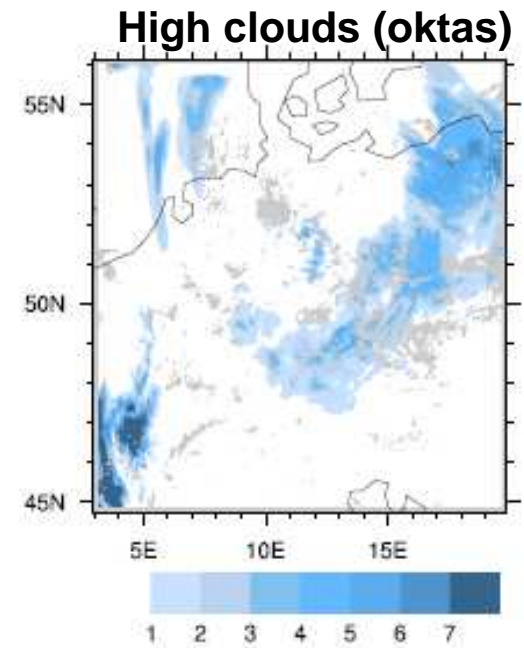
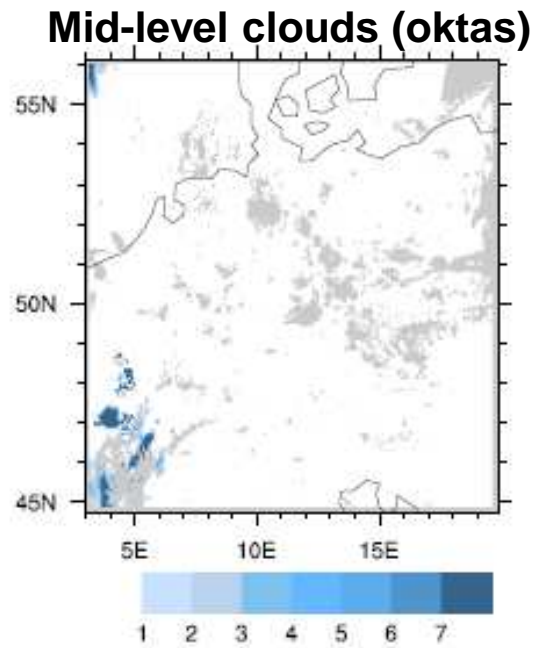
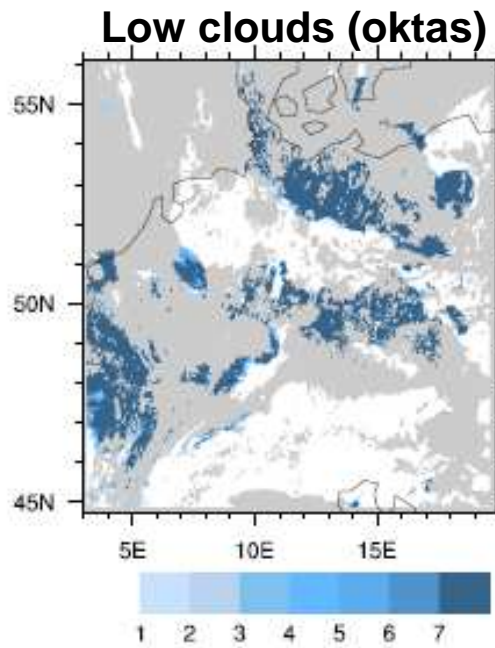


type of obs increment ,
for pixels observed to be **cloud-free** ?

- assimilate cloud fraction $CLC_{obs} = 0$ separately
for high, medium, low clouds
- model equivalent:
maximum CLC within vertical range

Use of CTH 'obs' for cloud-free pixels : Example

→ COSMO cloud fraction where observations “cloudfree”



17 Nov 2011, 6 UTC

Cycling LETKF data assimilation experiment : comparison

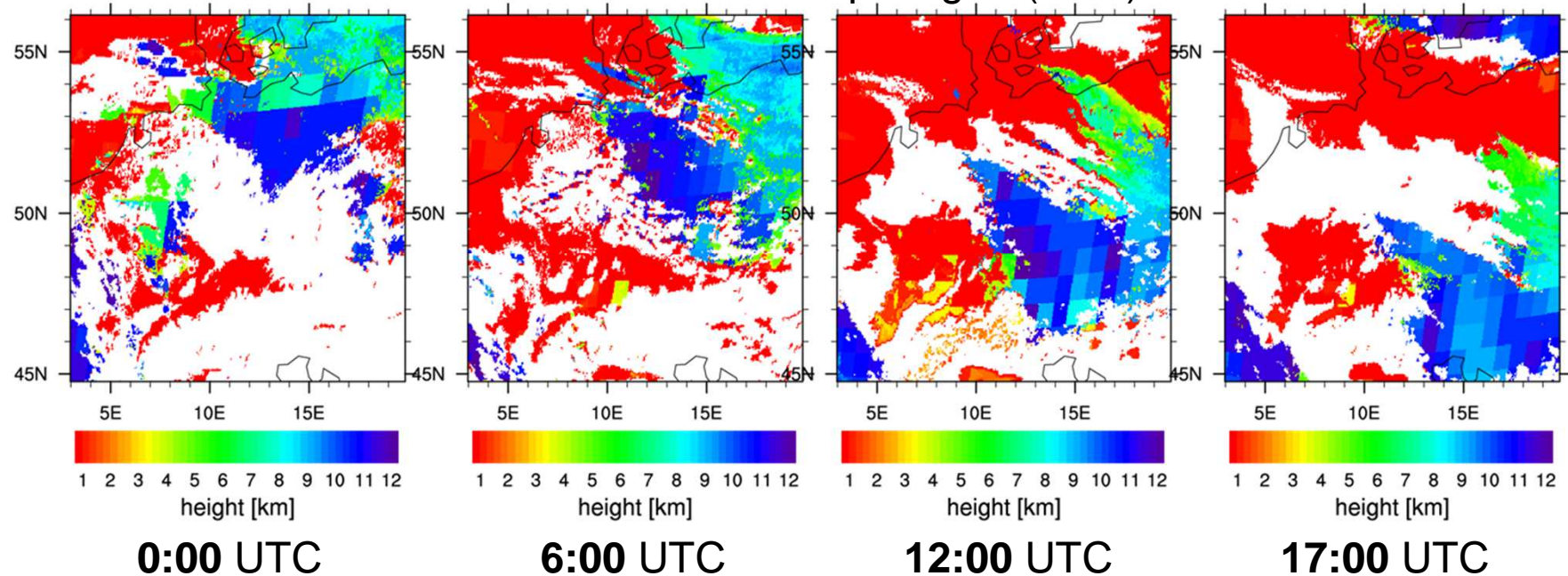
only conventional obs (radiosonde, aircraft, wind profiler, surface)

vs

conventional + cloud data

- LETKF: 40 ensemble members, obs thinning 14 km
- 1-hourly cycling over 21 hours, 13 Nov., 21 UTC – 14 Nov. 2011, 18 UTC
(wintertime low stratus)

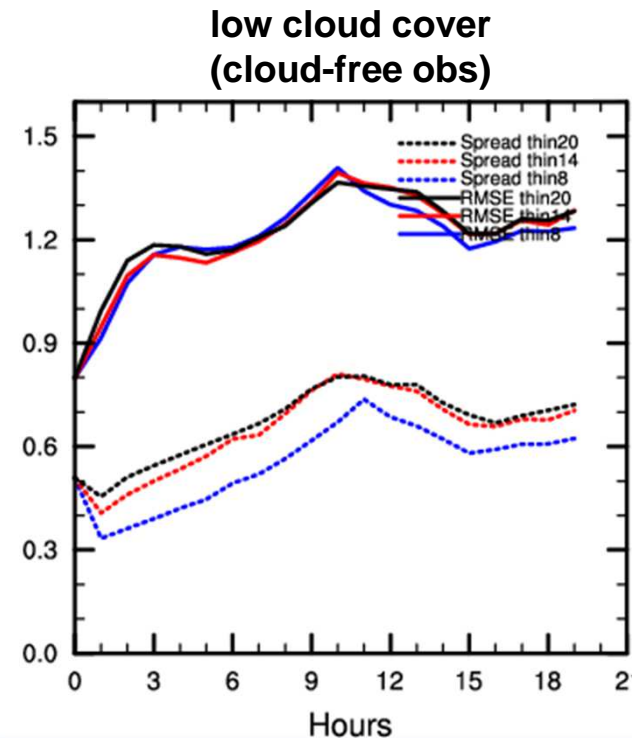
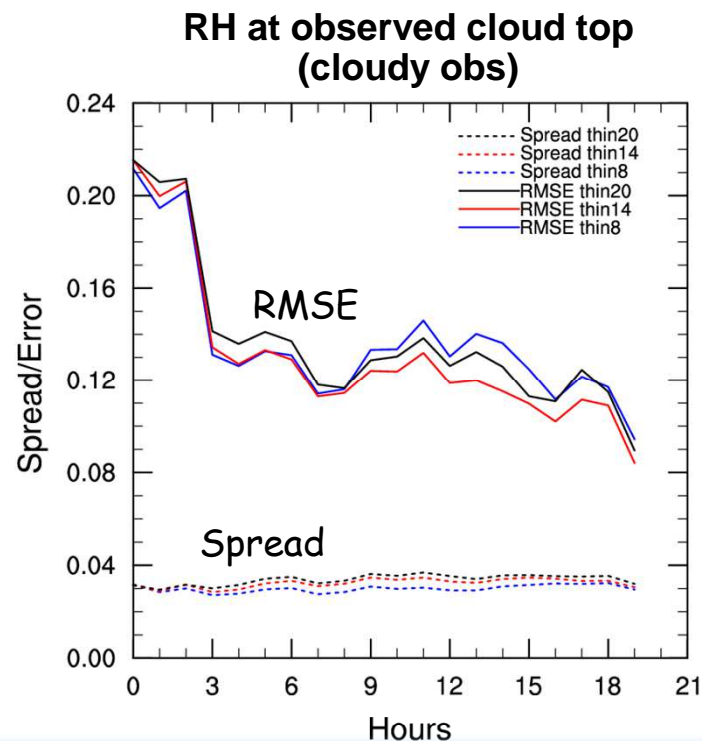
observed cloud top height (CTH)



Sensitivity experiment : data density

→ Comparing experiments with different data density:

- 8 km
- 14 km
- 20 km

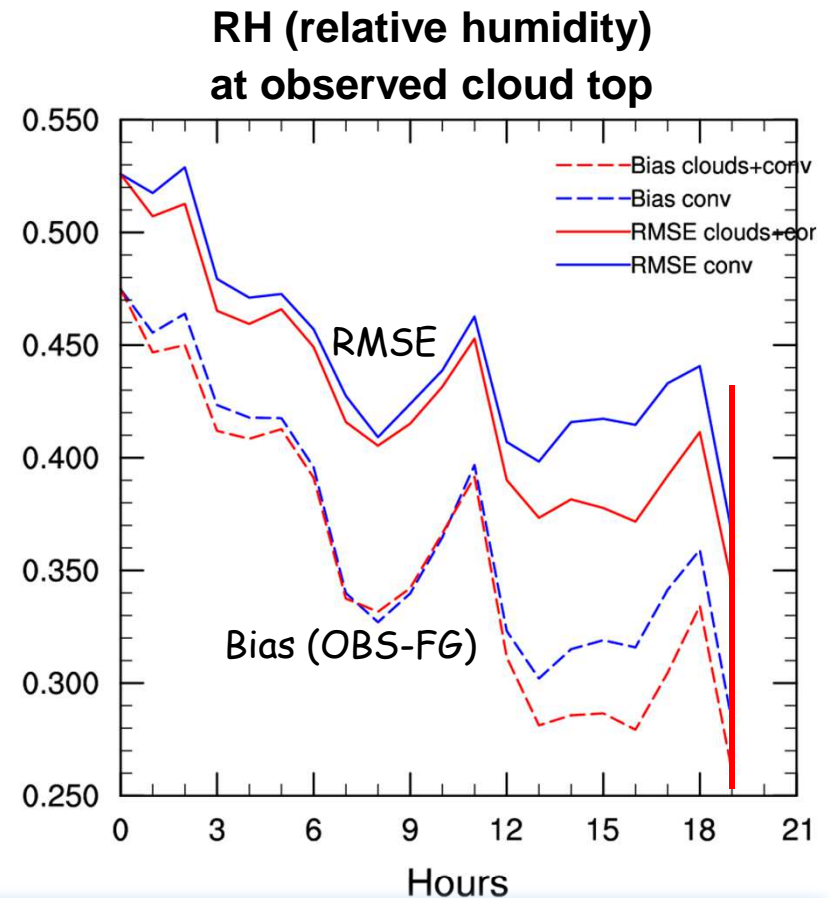


- *cloudy pixels: best results for a 14 km thinning*
- *cloud-free pixels: no clear conclusion*
- *lower spread for 8km thinning experiment (small difference in spread betw. 14 / 20km thinning)*
- *ensemble is underdispersive, but no sign of a further reduction of spread during the cycling*

Cycling LETKF: comparison 'only convective' versus 'conventional + cloud obs'

Time series of first guess errors, averaged over **cloudy** obs locations

assimilation of conventional obs only
assimilation of conventional + cloud obs

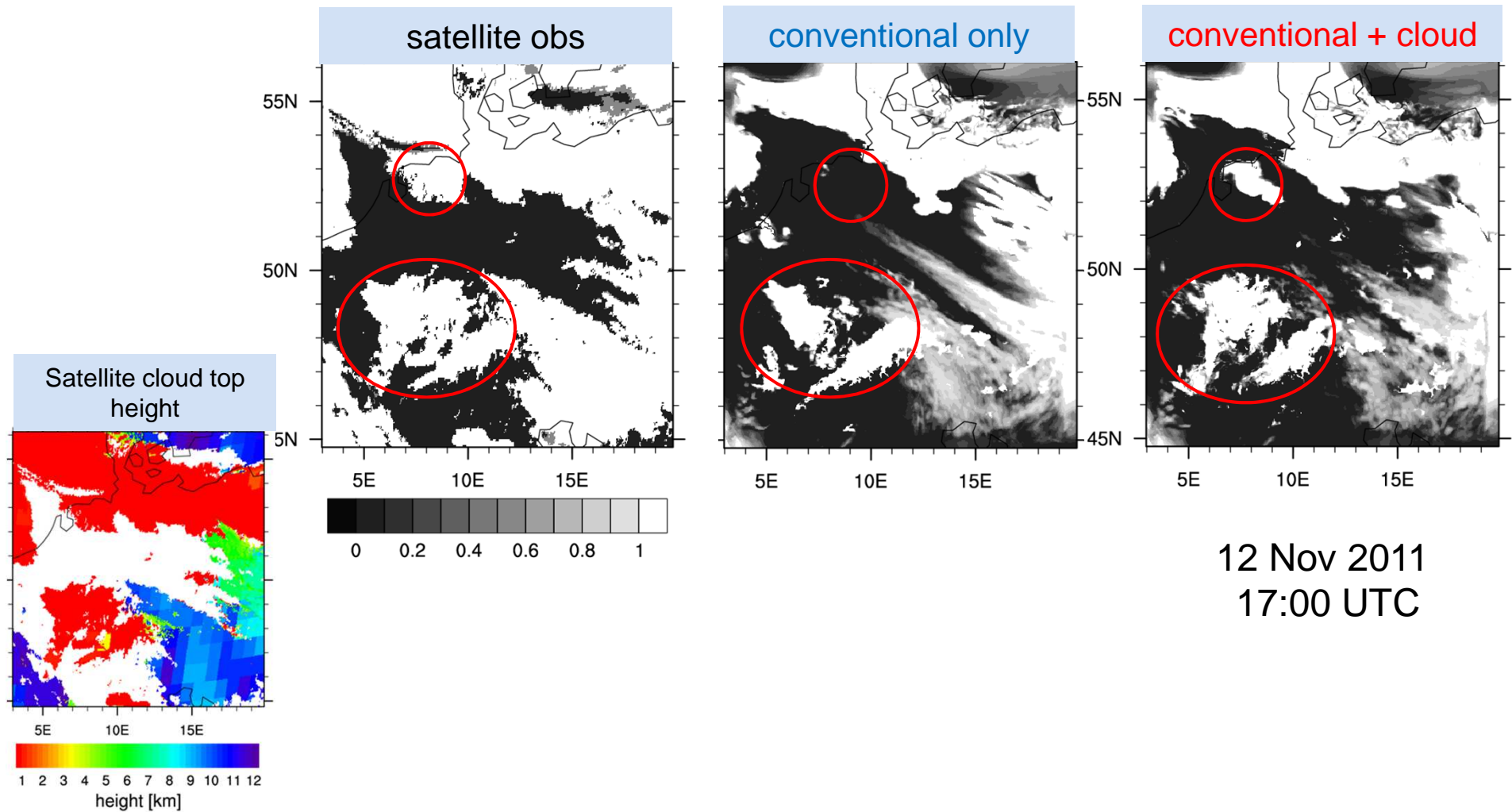


→ Cloud assimilation reduces RH (1-hour forecast) errors



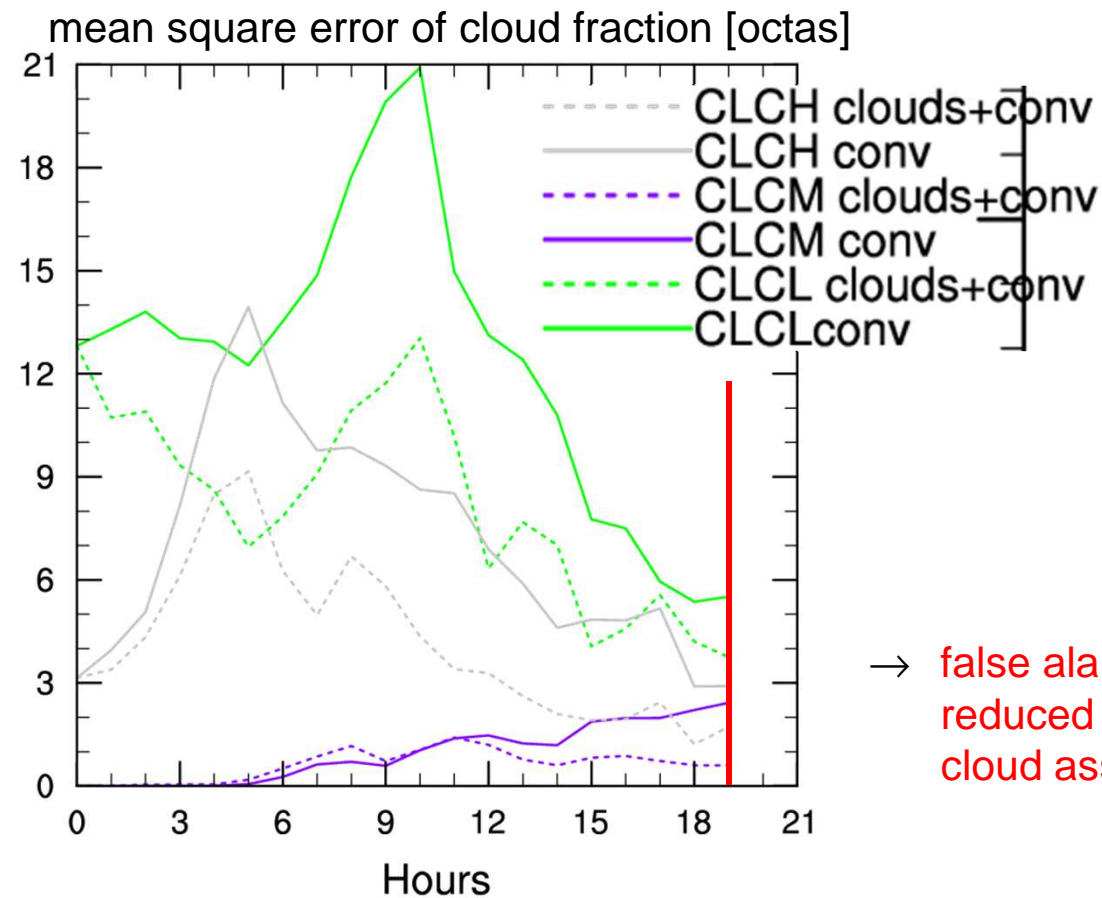
Cycling LETKF: comparison 'only convective' versus 'conventional + cloud obs'

Total cloud cover of first guess fields after 20 hours of cycling



Cycling LETKF: comparison 'only convective' versus 'conventional + cloud obs'

time series of first guess errors, averaged over **cloud-free** obs locations
(errors are due to false alarm cloud)



→ false alarm clouds reduced through cloud assimilation

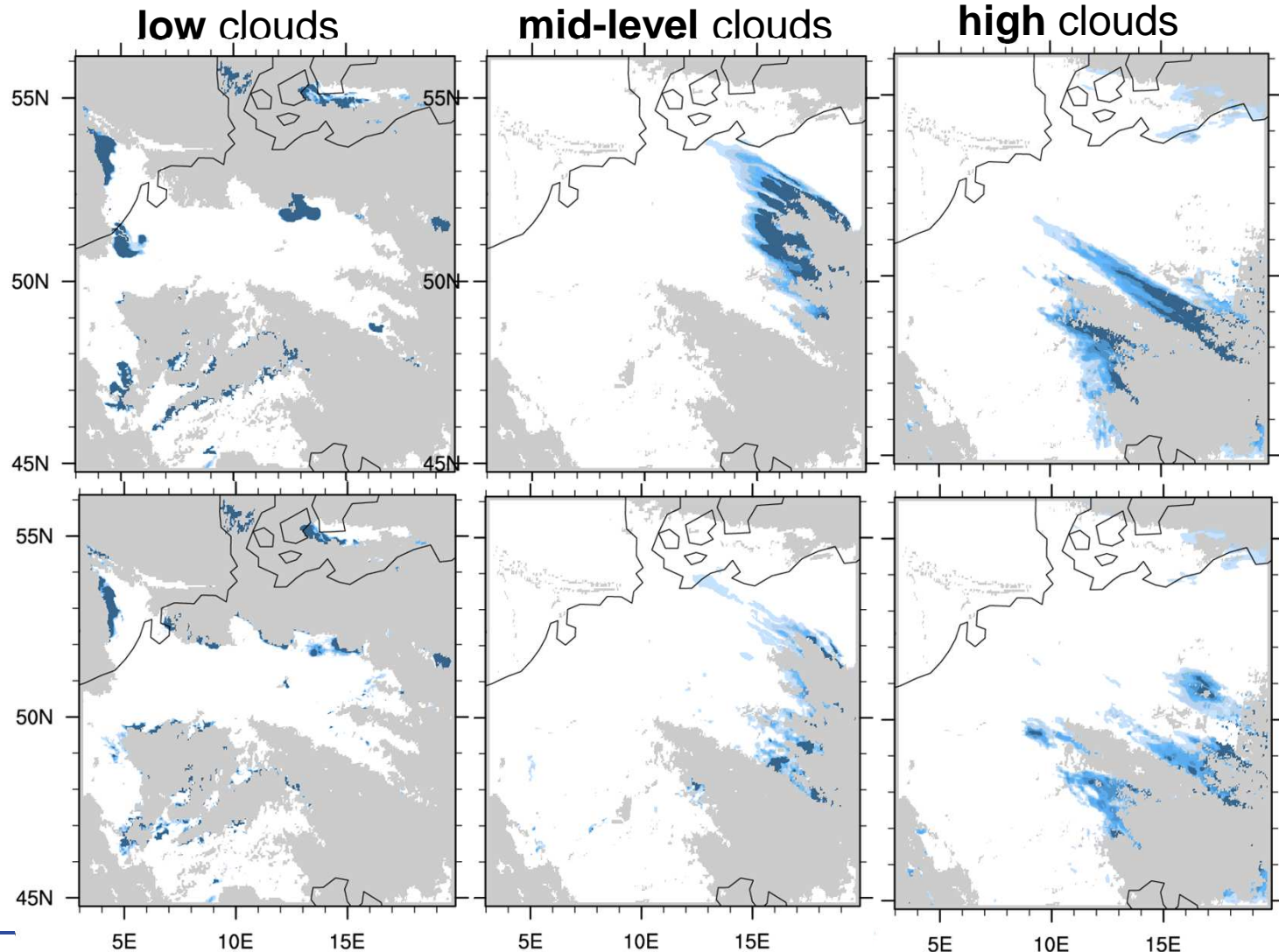
Cycling LETKF: comparison 'only convective' versus 'conventional + cloud obs'

'false alarm'
cloud cover
(after 20 hrs cycling)

conventional
obs only

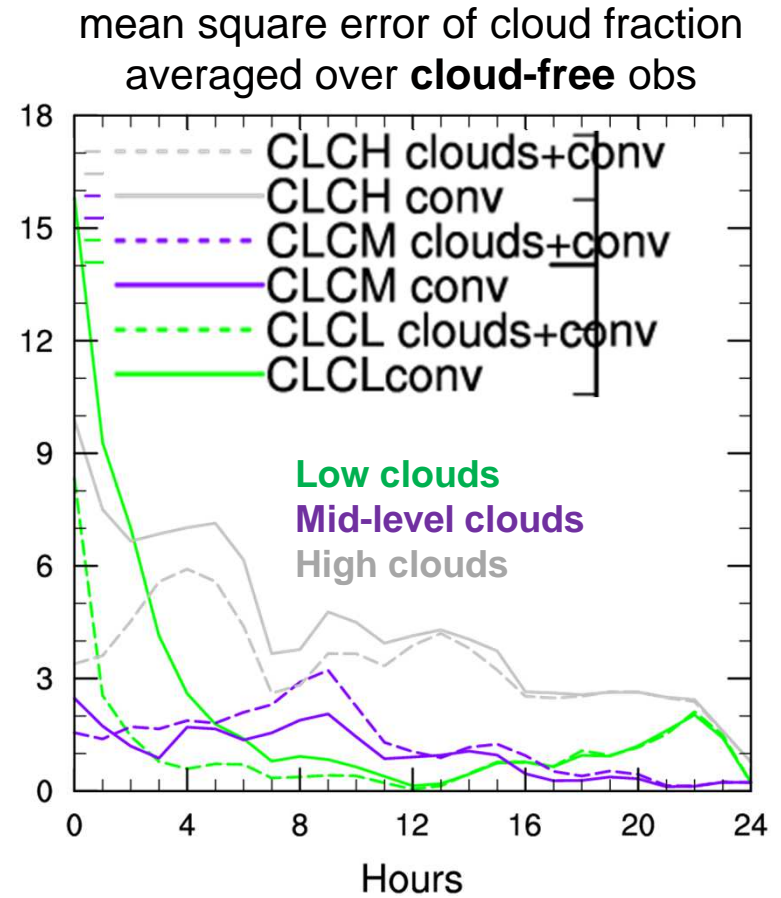
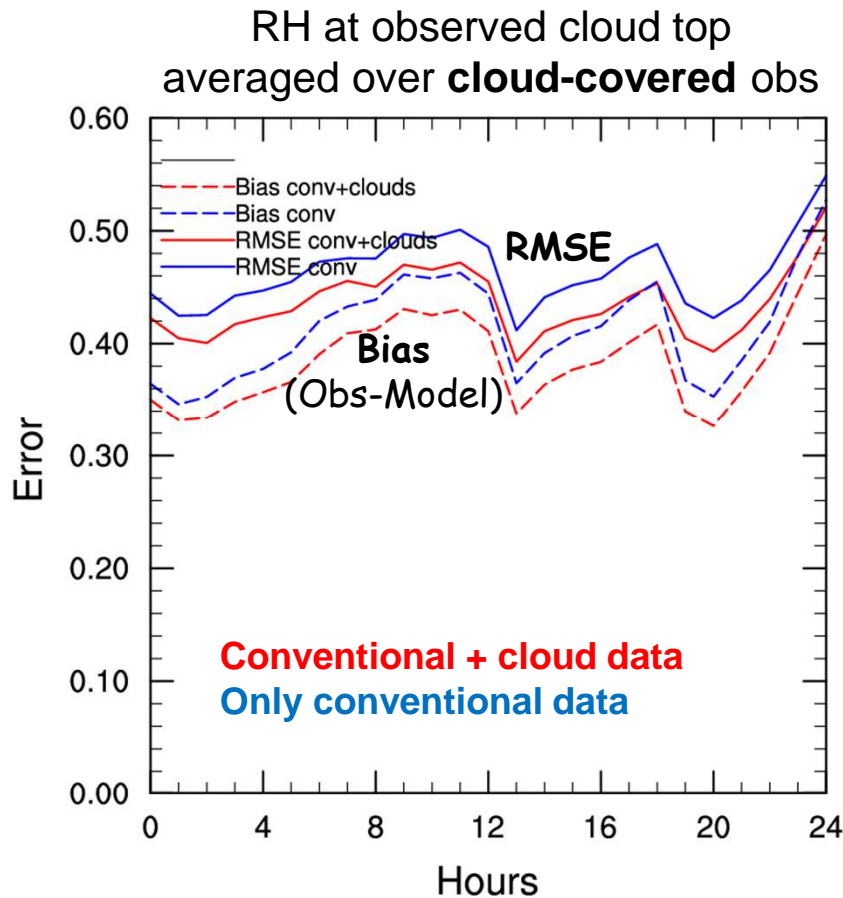


conventional
+ cloud



Comparison of free deterministic forecast (after 12 hrs DA): time series of errors

(forecast starts 14 Nov., 9 UTC)



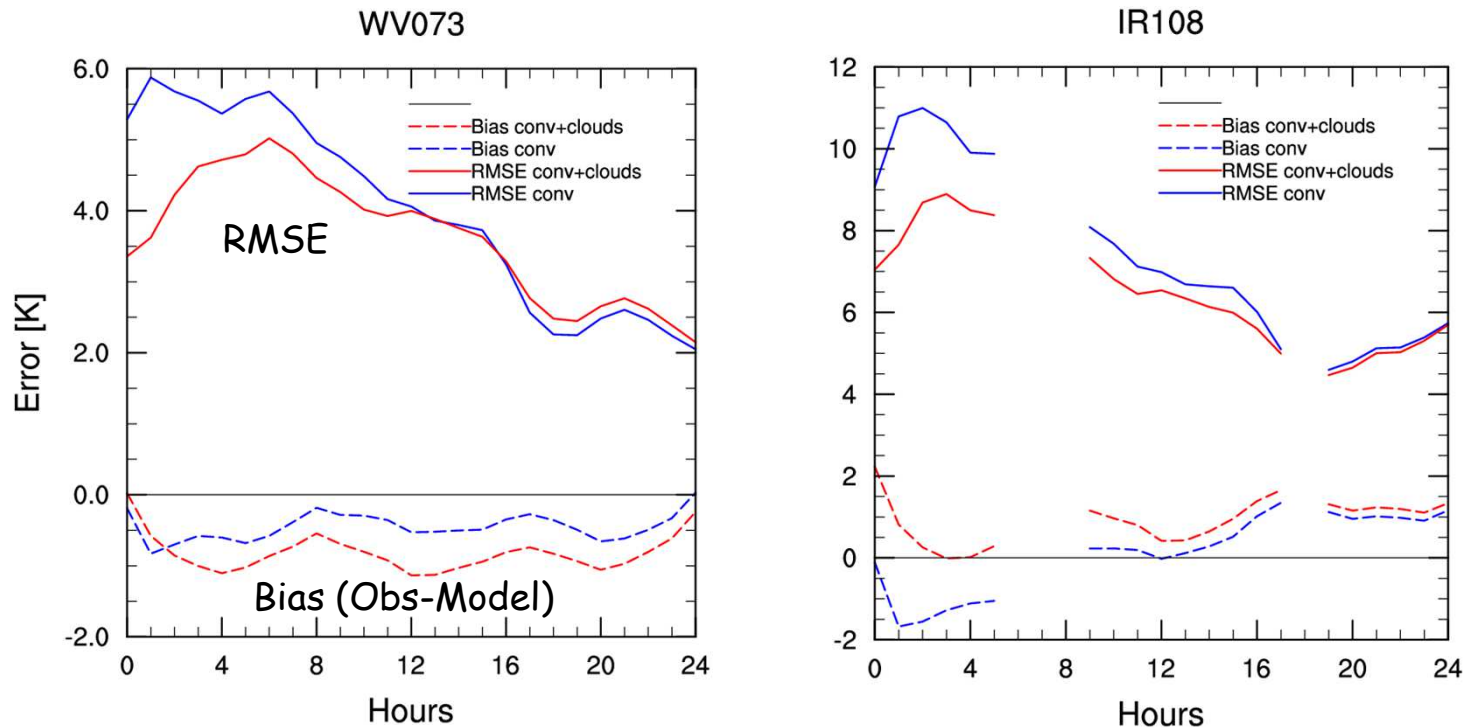
➔ **The forecast of cloud characteristics can be improved through the assimilation of the cloud information**



Verification of free deterministic forecast against independent observations

Errors for SEVIRI infrared brightness temperatures
(model values computed with RTTOV)

Conventional + cloud data
Only conventional data



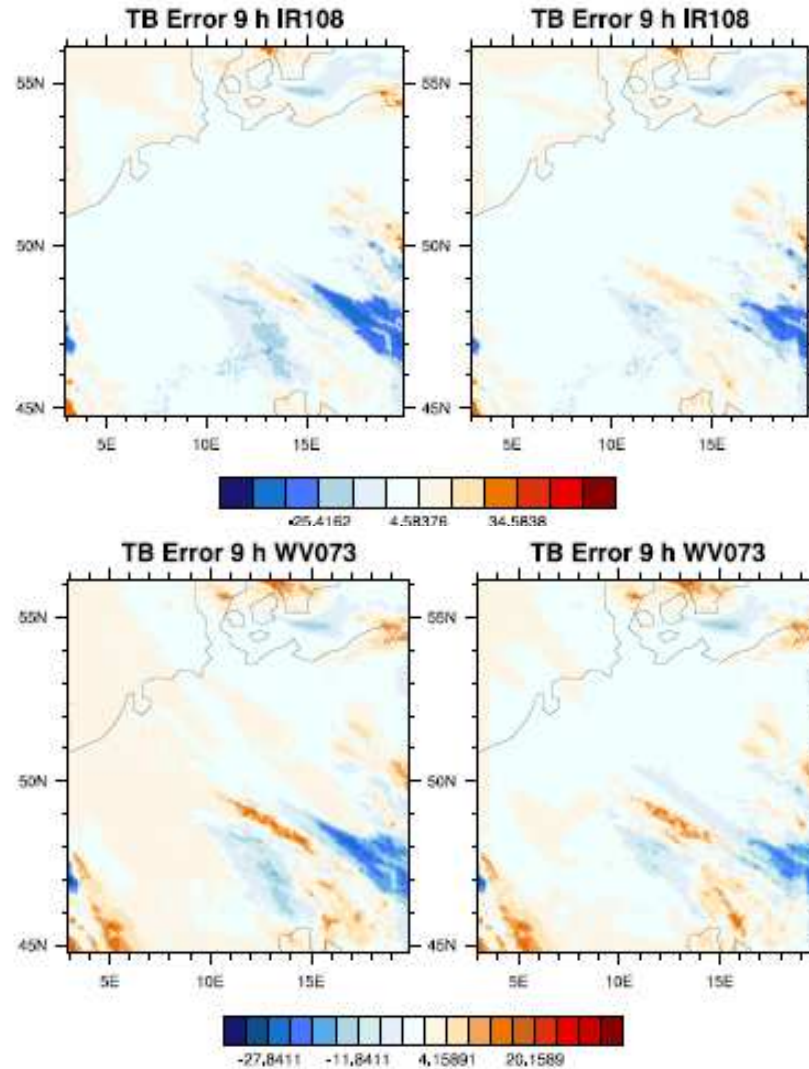
→ RMSE is smaller for first 16 hours of forecast for cloud experiment, bias varies



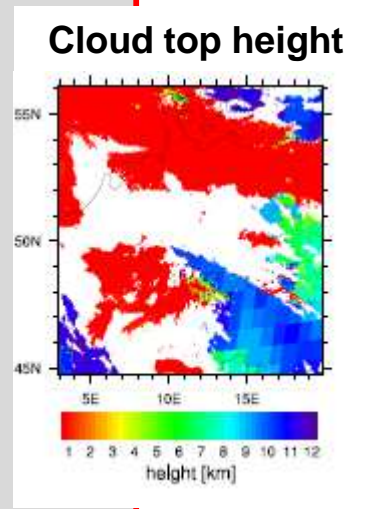
Verification of 9-h forecast against indep. obs: SEVIRI brightness temperature

Only CONV
experiment

14 Nov 2011,
18 UTC




CONV+CLOUD
experiment



→ Also the high clouds are simulated better in the cloud experiment

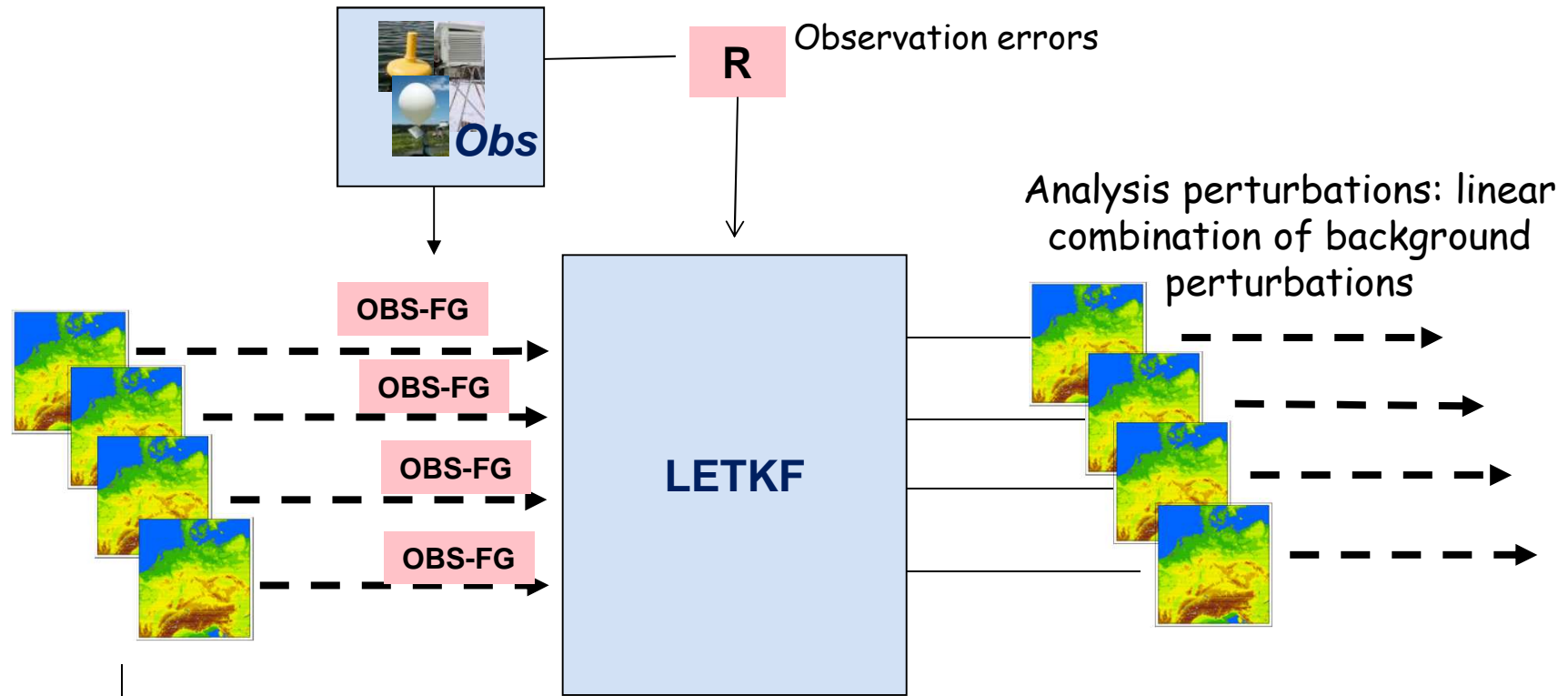
Use of (SEVIRI-based) cloud observations in LETKF:

- Tends to **introduce humidity / cloud where** it should
- Tends to **reduce 'false-alarm' clouds**
- **Improvement on cloud characteristics in free forecast** for a stable wintertime high-pressure systems
- May also be useful for **convective** situations
 - If convective clouds are captured better while developing, convective precipitation may be improved → needs to be tested

- Evaluate other variables and other cases, longer time series
- Application in project EWeLiNE: Improving the forecast for renewable energy sector (clouds particularly important for photovoltaic power production)

- Also work on direct SEVIRI radiance assimilation
(together with Africa Perianez, Robin Faulwetter)

Thank you for your attention!

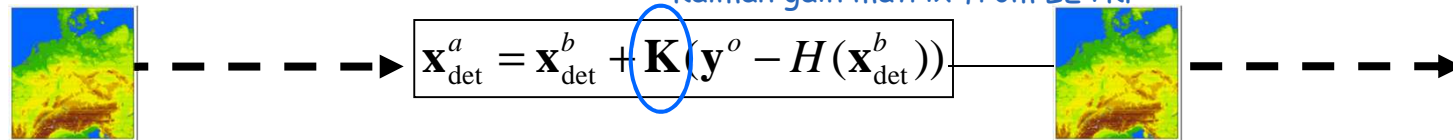
Local Ensemble Transform Kalman Filter



Background error correlations

$$\mathbf{P}^b = \frac{1}{(k-1)} \sum_{i=1}^k (\mathbf{x}^{b(i)} - \bar{\mathbf{x}}^b)(\mathbf{x}^{b(i)} - \bar{\mathbf{x}}^b)^T$$

Additional: one deterministic run:



The diagram shows a deterministic run where a background field (weather map) is updated to an analysis field (weather map) using the Kalman gain matrix from LETKF. The equation is:

$$\mathbf{x}_{\text{det}}^a = \mathbf{x}_{\text{det}}^b + \mathbf{K}(\mathbf{y}^o - H(\mathbf{x}_{\text{det}}^b))$$

The Kalman gain matrix \mathbf{K} is circled in blue in the original image.