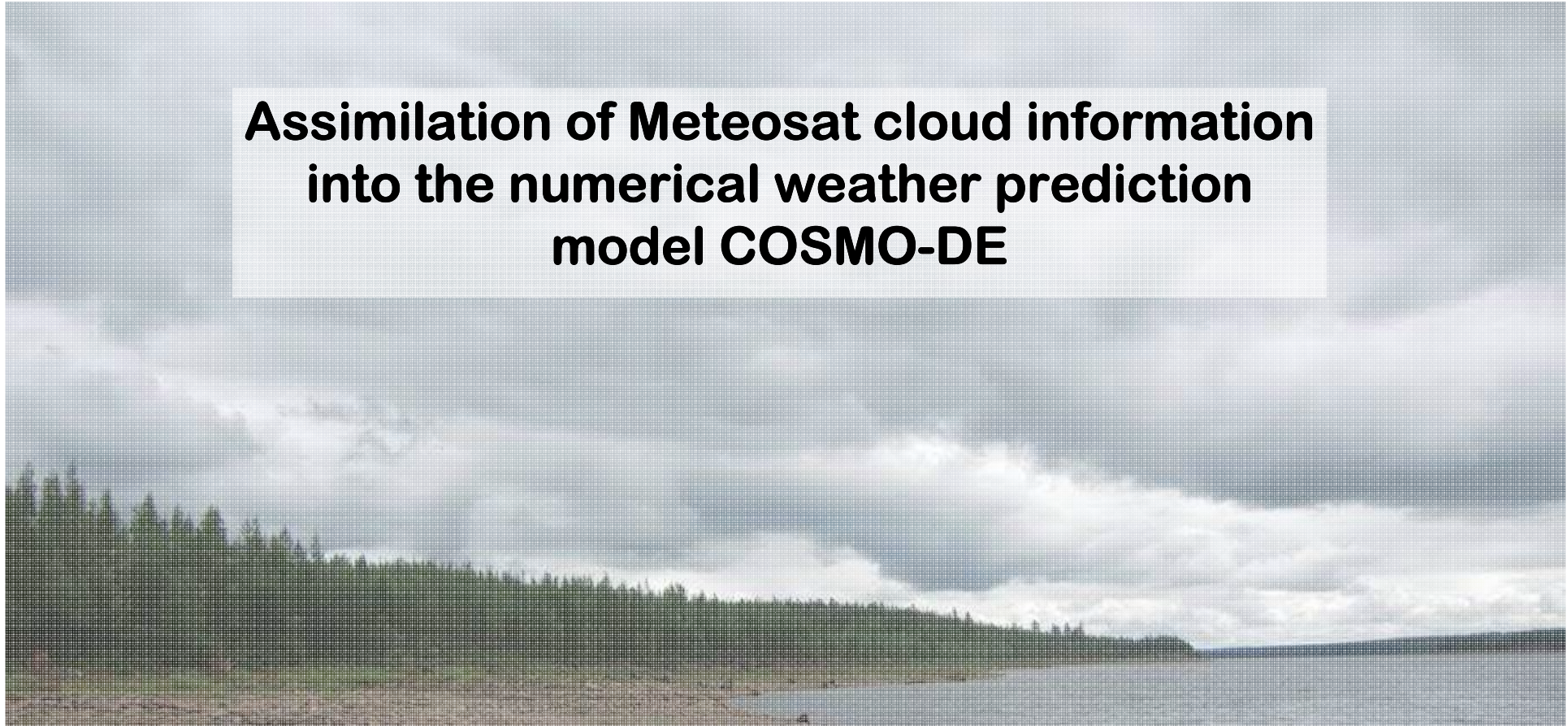
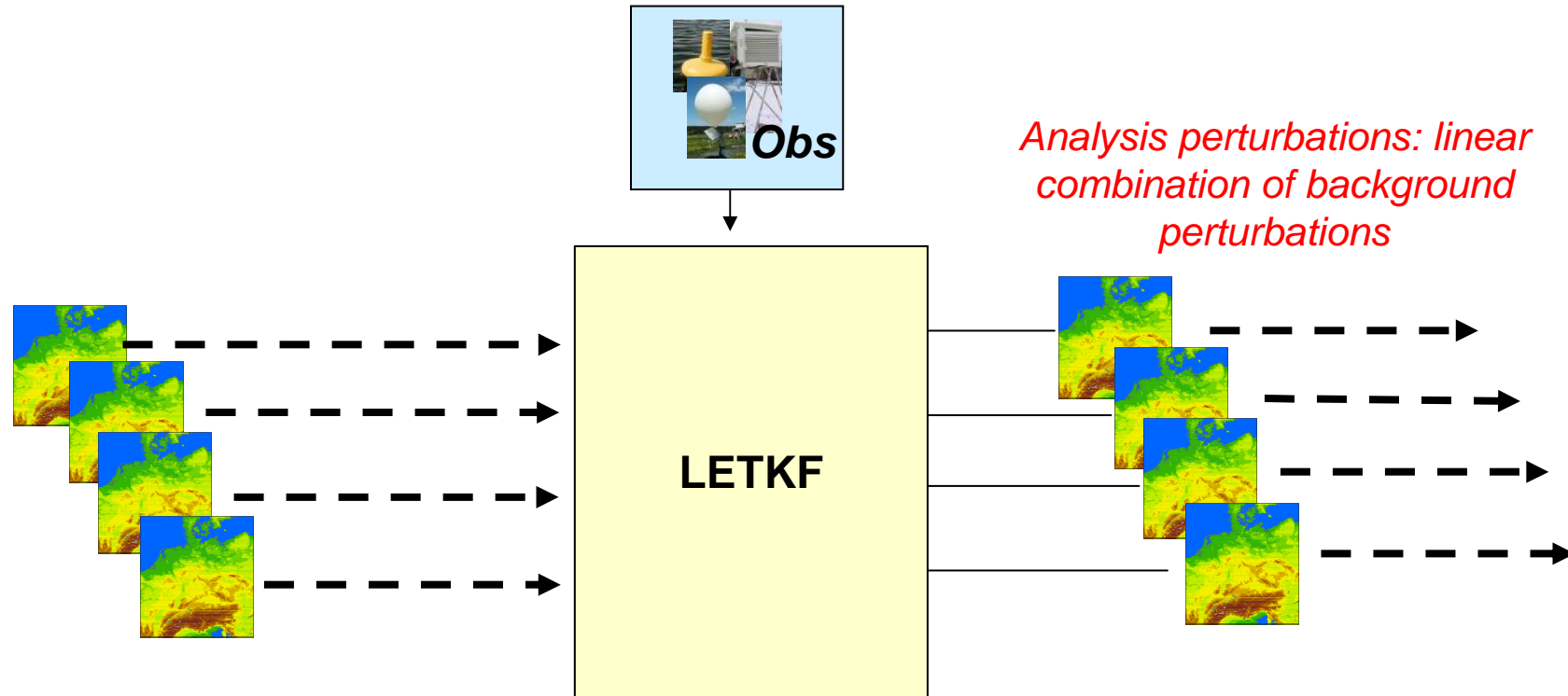


Assimilation of Meteosat cloud information into the numerical weather prediction model COSMO-DE

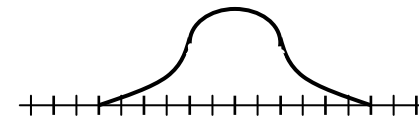


Annika Schomburg, Christoph Schraff

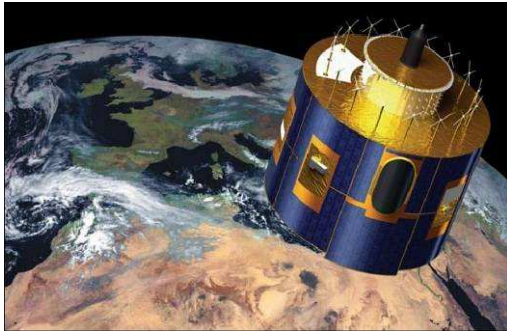
Local Ensemble Transform Kalman Filter in COSMO (KENDA project)



- **L**ocal: the linear combination is fitted in a local region
 - observation have a spatially limited influence region



Assimilation of cloud information into COSMO-DE

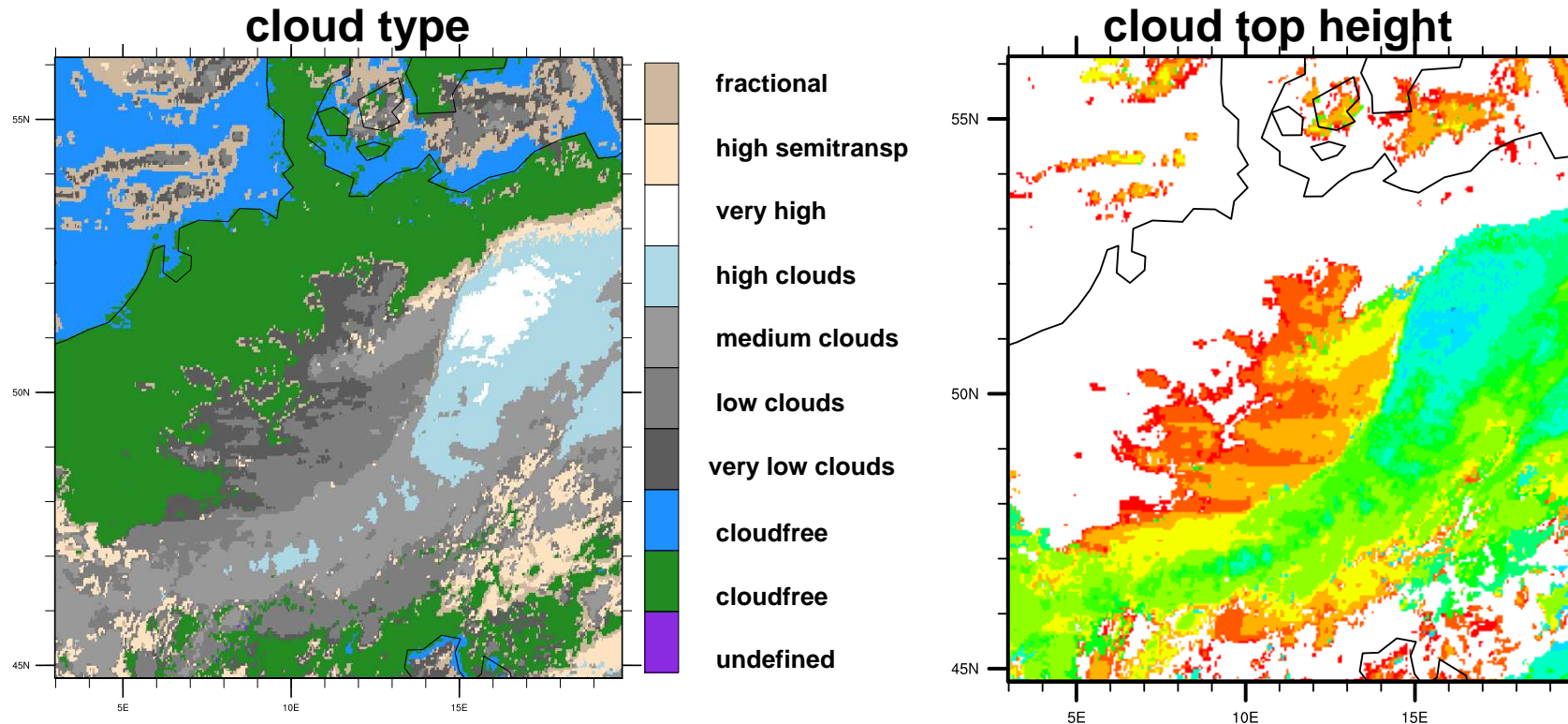


Source: EUMETSAT

- NWCSAF cloud products based on satellite data:
 - **Meteosat-SEVIRI**
 - $\Delta x \sim 5\text{km}$ over central Europe,
 - $\Delta t = 15\text{ min}$

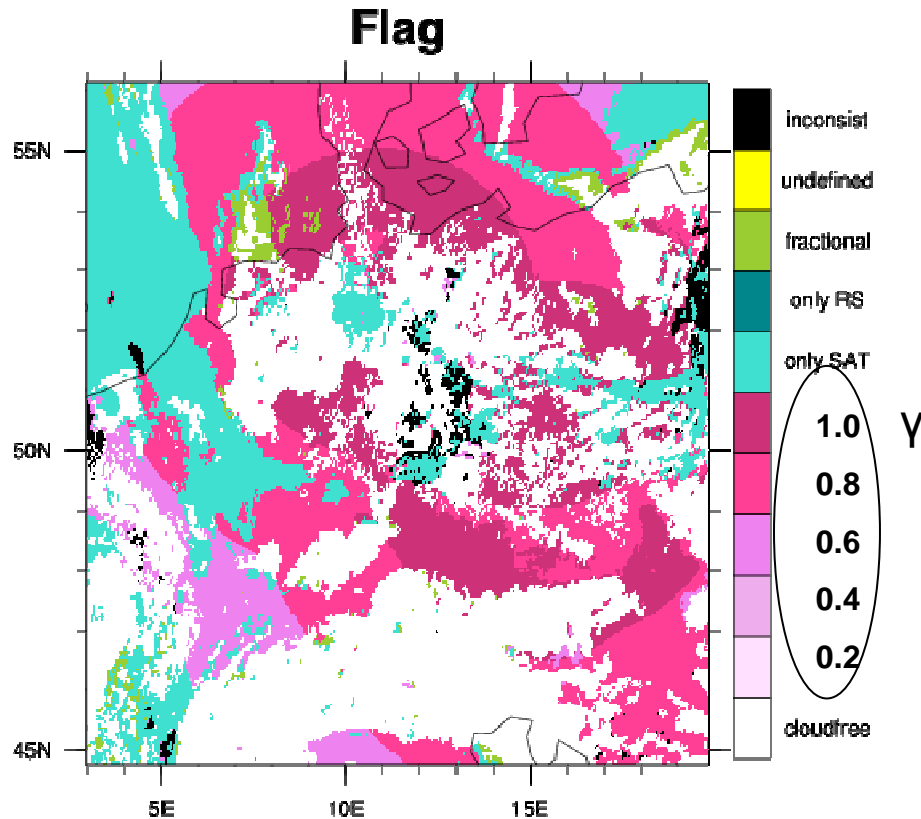
Satellite retrievals: NWCSAF cloud products (SEVIRI/MSG)

Deutscher Wetterdienst
Wetter und Klima aus einer Hand



Retrieval algorithm needs temperature and humidity profile information from a NWP model → *cloud top height might be at wrong height if temperature-profile in NWP model is not simulated correctly!* → use also radiosonde information where available

Combine satellite & radiosonde information: data availability flag



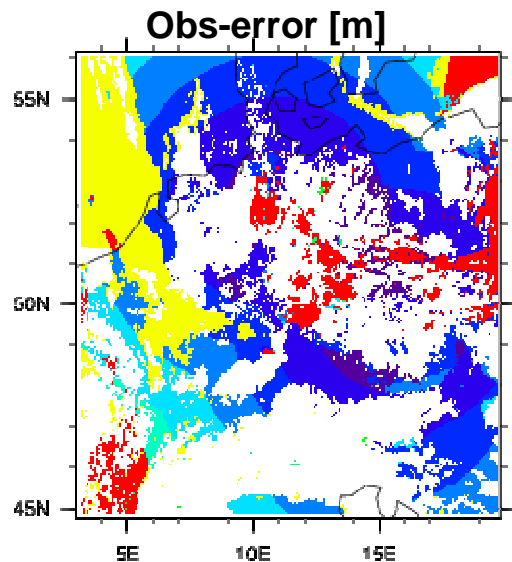
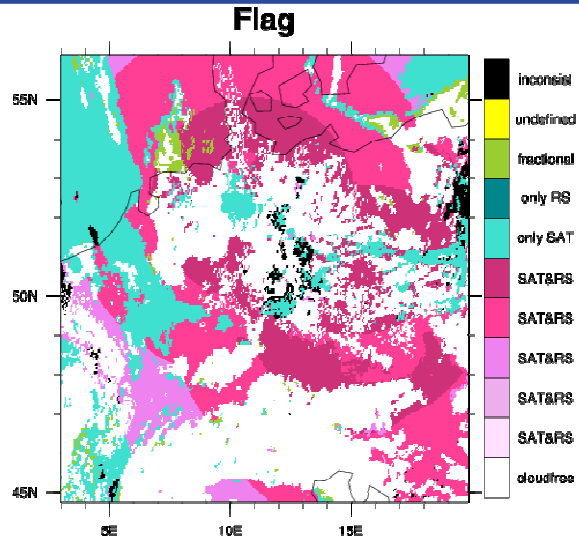
- Use temporal and spatial distance of radiosonde for weighting:

$$cth_{corr} = (1 - \gamma)cth_{sat} + \gamma \cdot cth_{rs}$$

- Also use data availability flag for observation error specification:

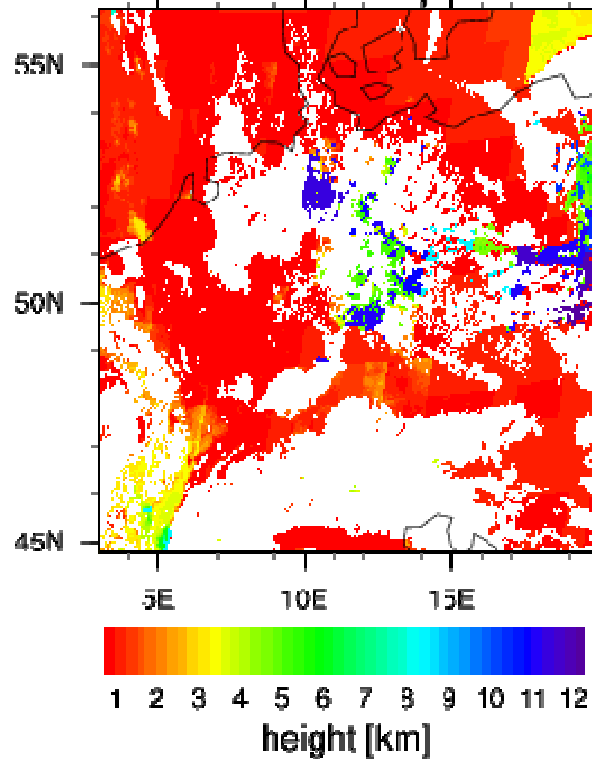
$$e_o = (1 - \gamma)e_{sat} + \gamma \cdot e_{rs}$$

Combine satellite & radiosonde information

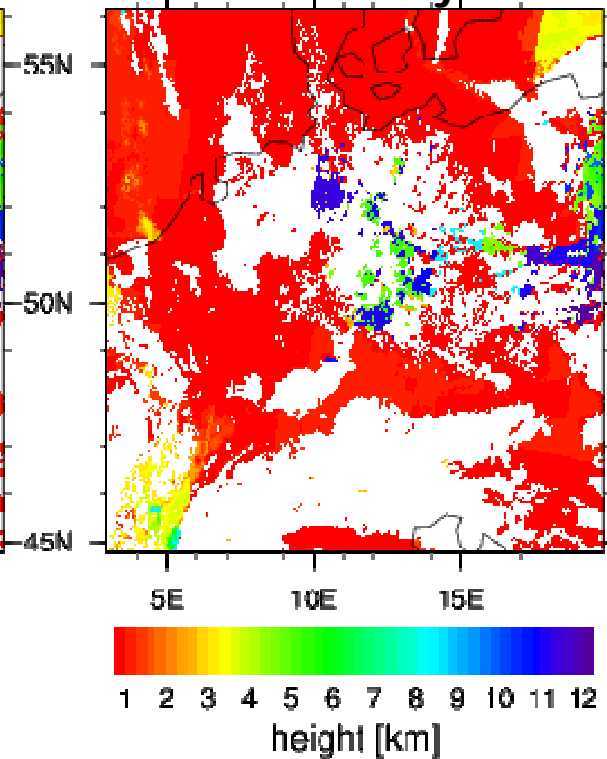


Cloud top height

Satellite cloud product

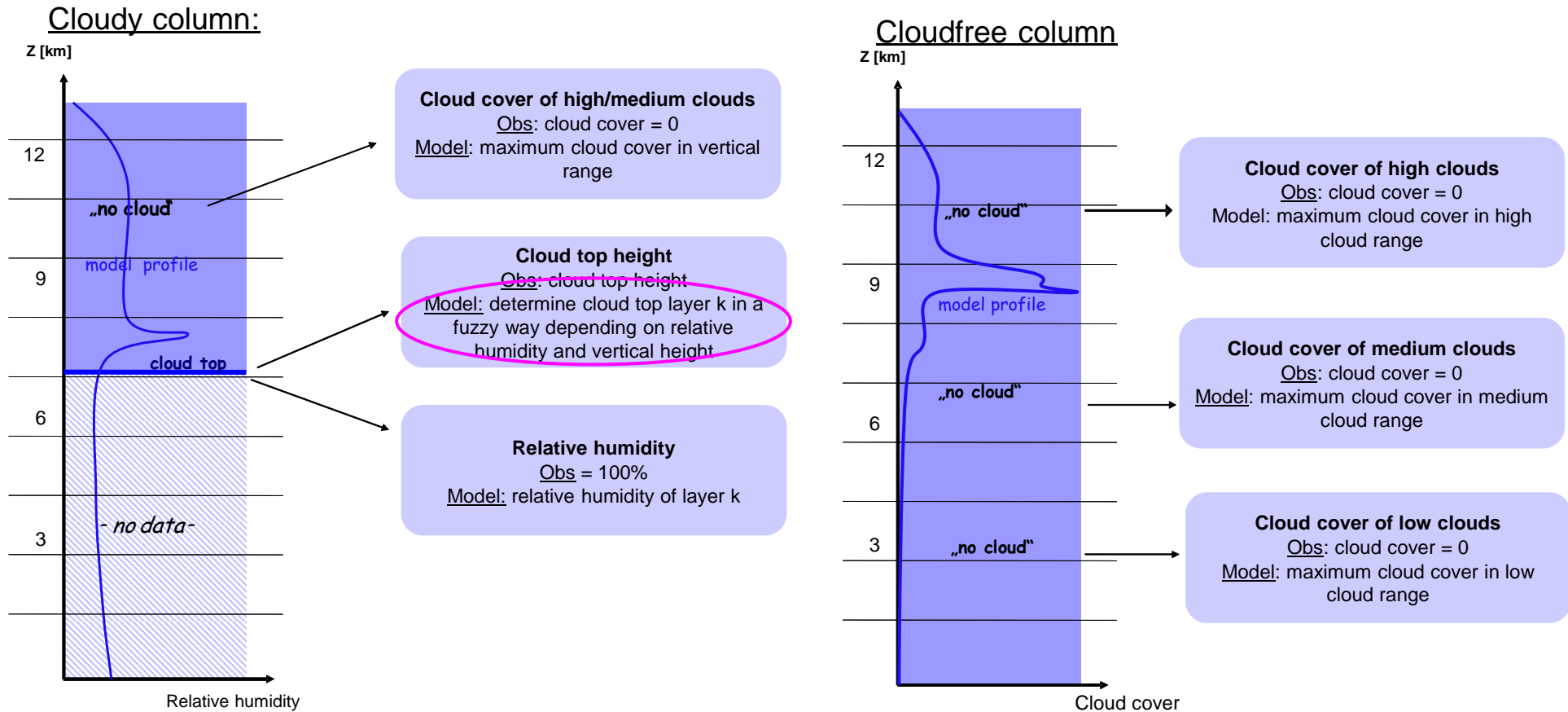


Cloud analysis

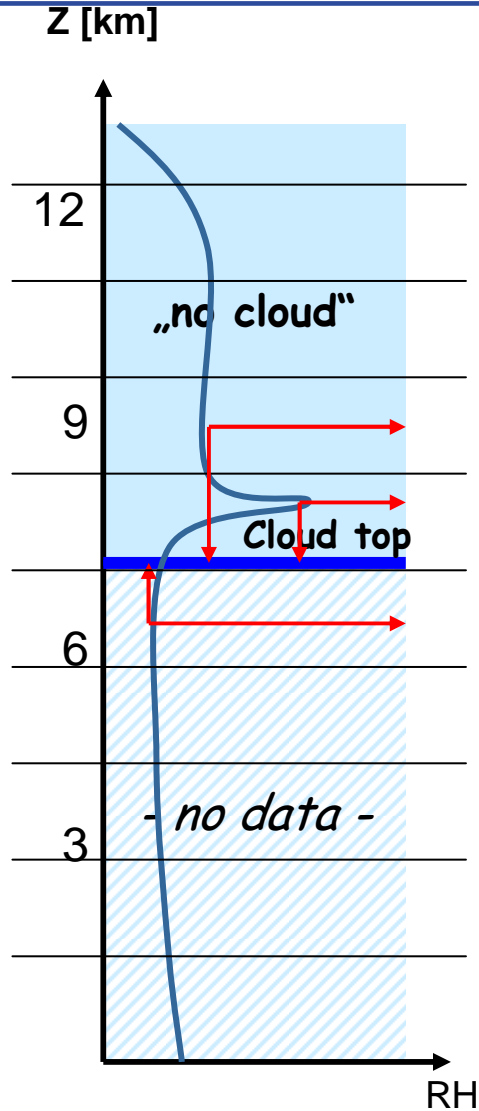


Variables assimilated

From one observation of cloud top height **several variables** are extracted and used to weight the ensemble members in the LETKF:



Find cloud top height model equivalent



- If using a fixed threshold to define cloud top, one might penalize close members
- Therefore: **find model layer optimally fitting the observed cloud top height:**

$$d = \min_k \sqrt{(f(\rho_k) - f(\rho_o))^2 + \frac{1}{\Delta h_{\max}} (h_k - h_o)^2}$$

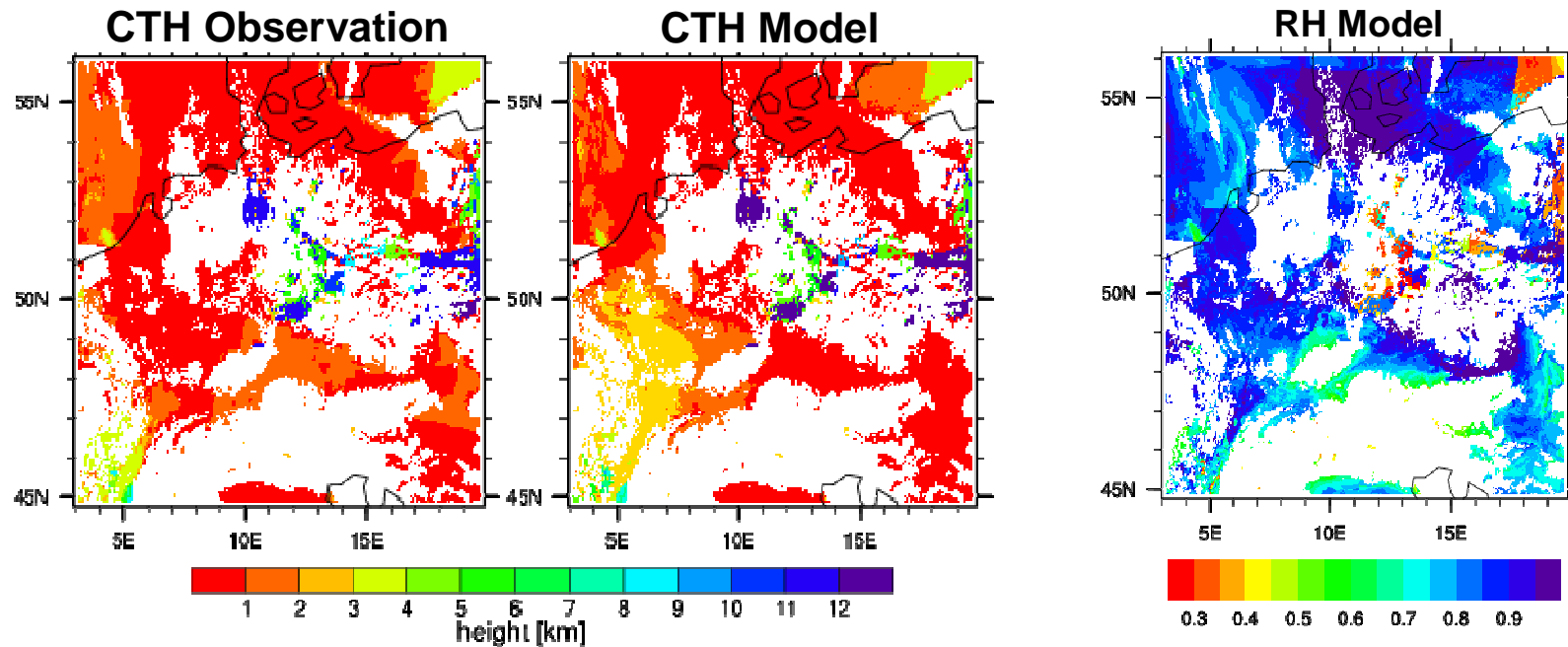
ρ : Relative humidity

h : height

- Search in a vertical range (e.g. +/-2500m) the minimum
- If above a layer exceeds the cloud coverage of the chosen layer or exceeds 70%, then chose the top of that layer

Model equivalents for cloudy column

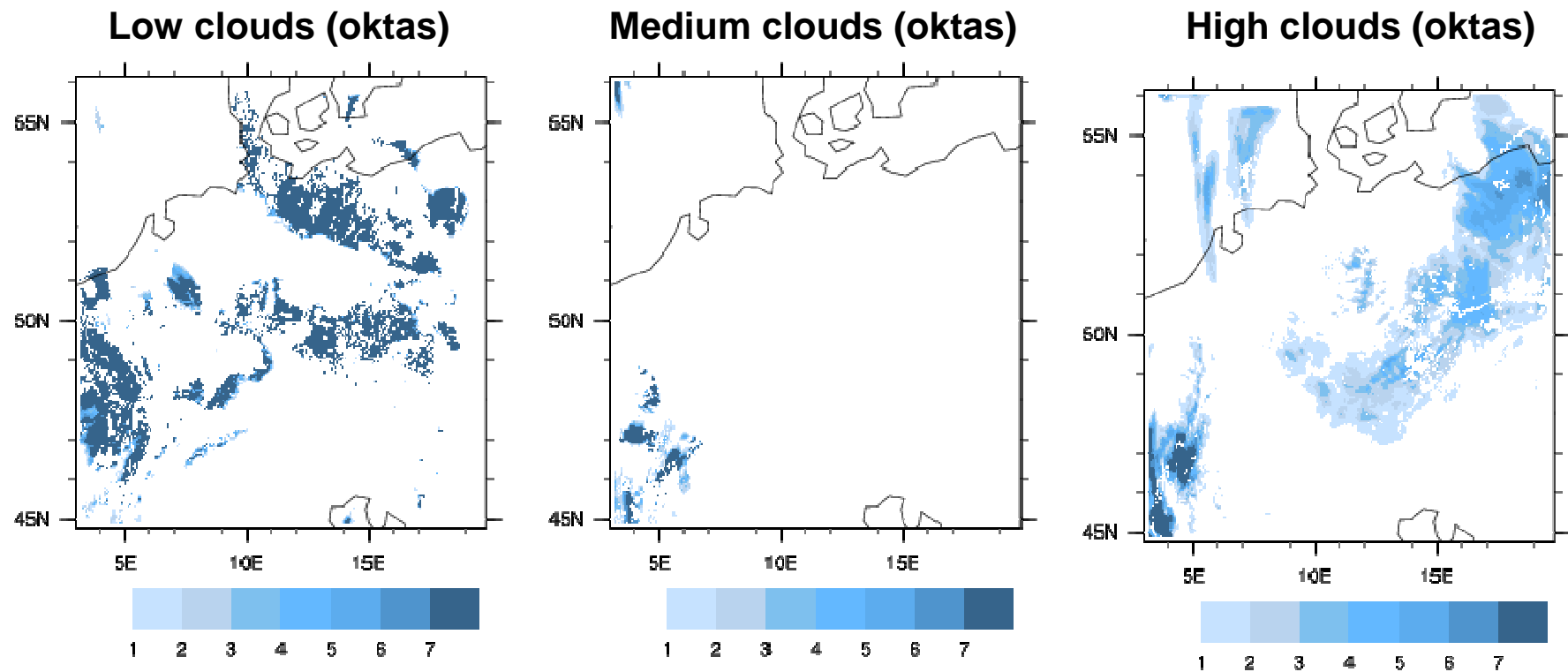
Assimilated variables: Cloud top height and relative humidity



Model equivalents for cloud-free column

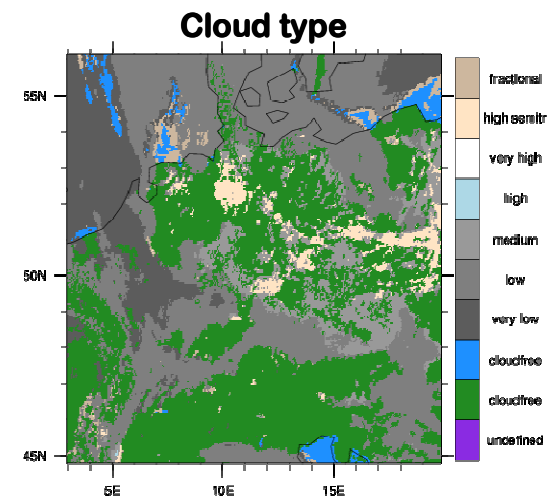
Assimilated variables: Cloud cover

- COSMO cloud cover where observations “**cloudfree**”



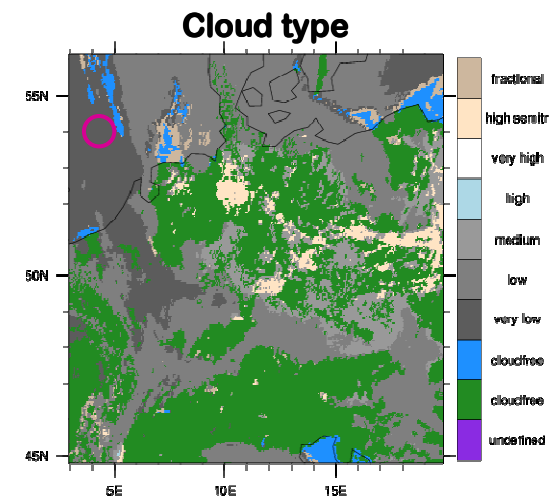
„Single observation“ experiment

- Analysis for 17 November 2011, 6:00 UTC (no cycling)
 - 32 member
- Assimilate every 60th column
- Horizontal localization: 20km
(weights equal to zero at ~73km=26 grid points)
- Objective:
 - Understand in detail what the filter does with such special observation types
 - Does it work at all?
 - Sensitivity to settings

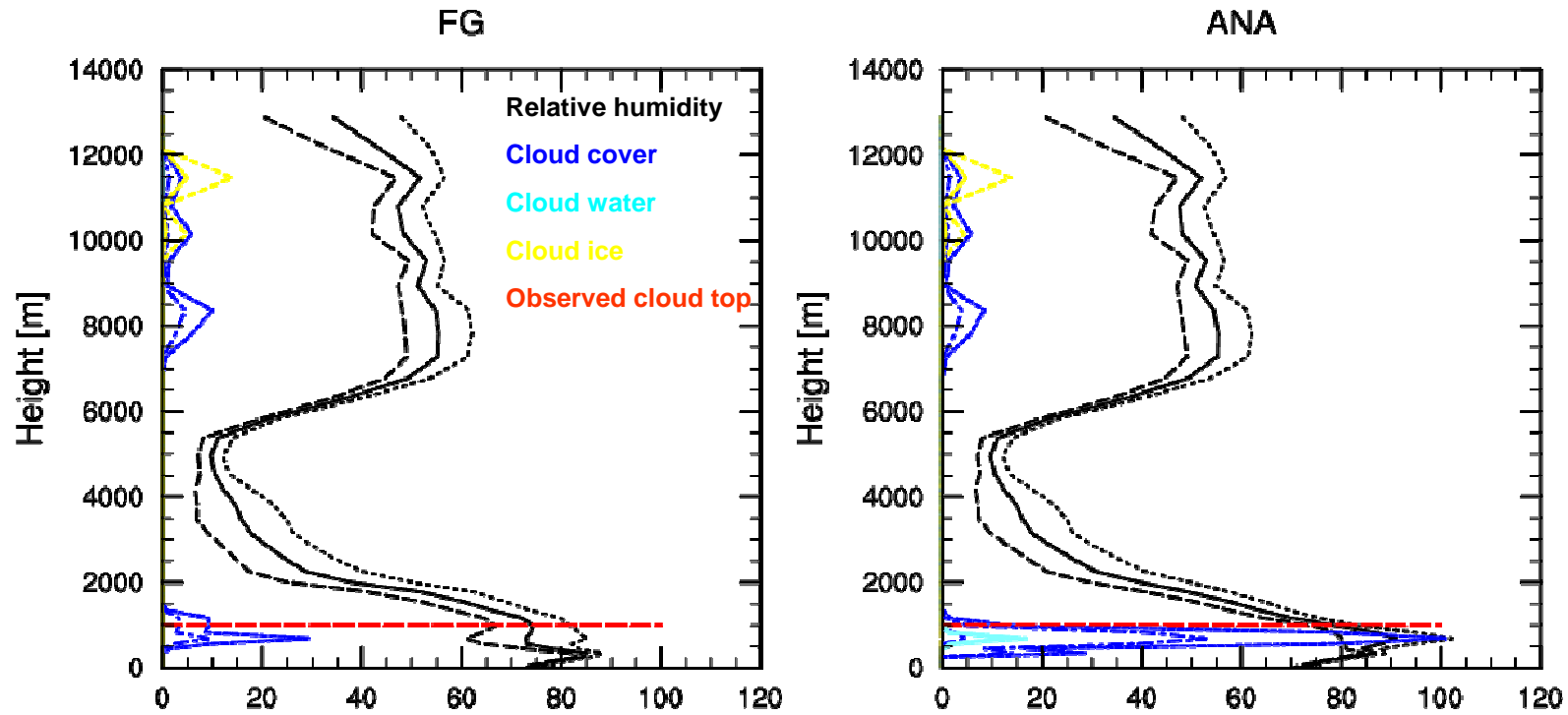


Example 1: no cloud in model

lat 53.9, lon 4.1

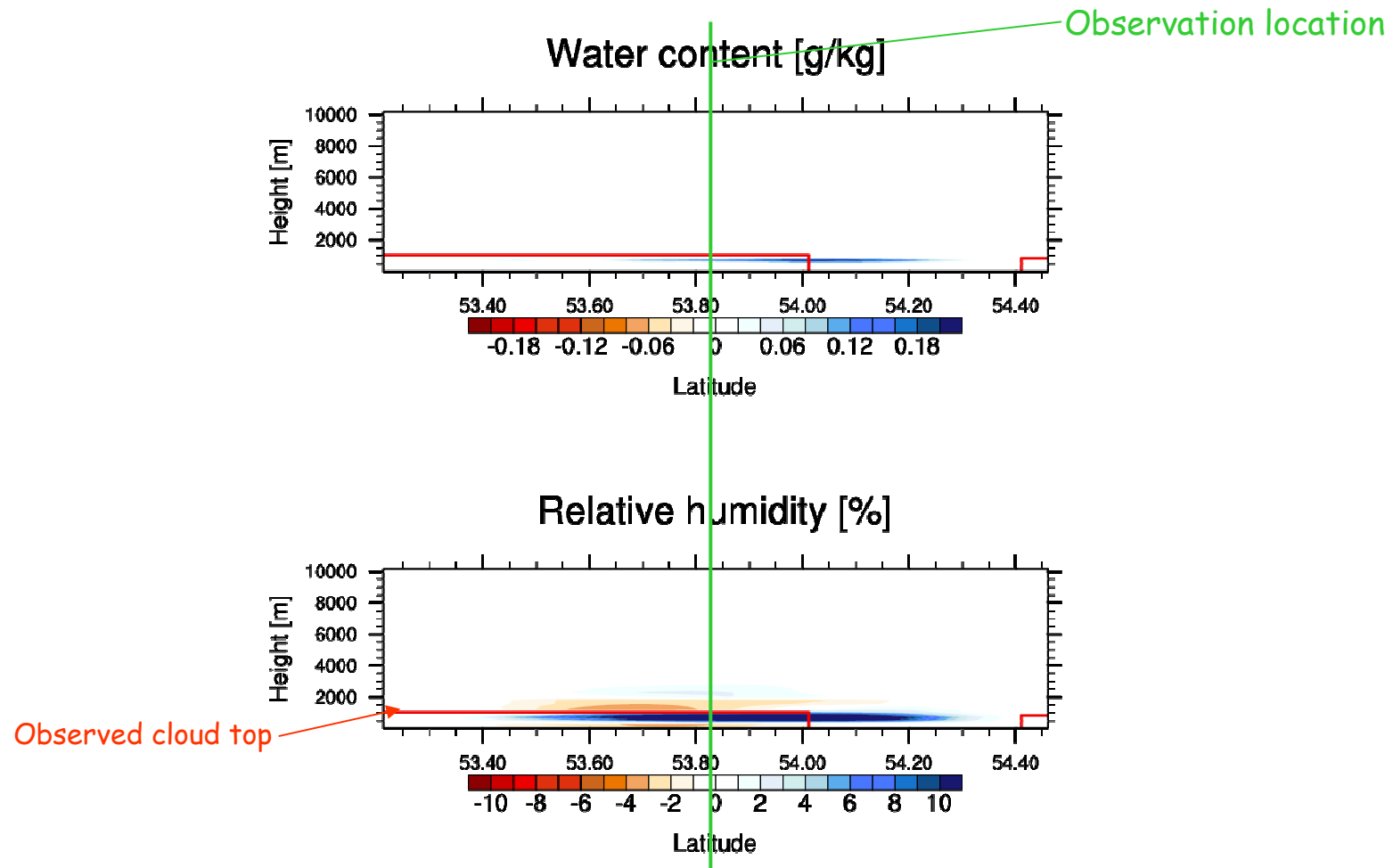


Profiles

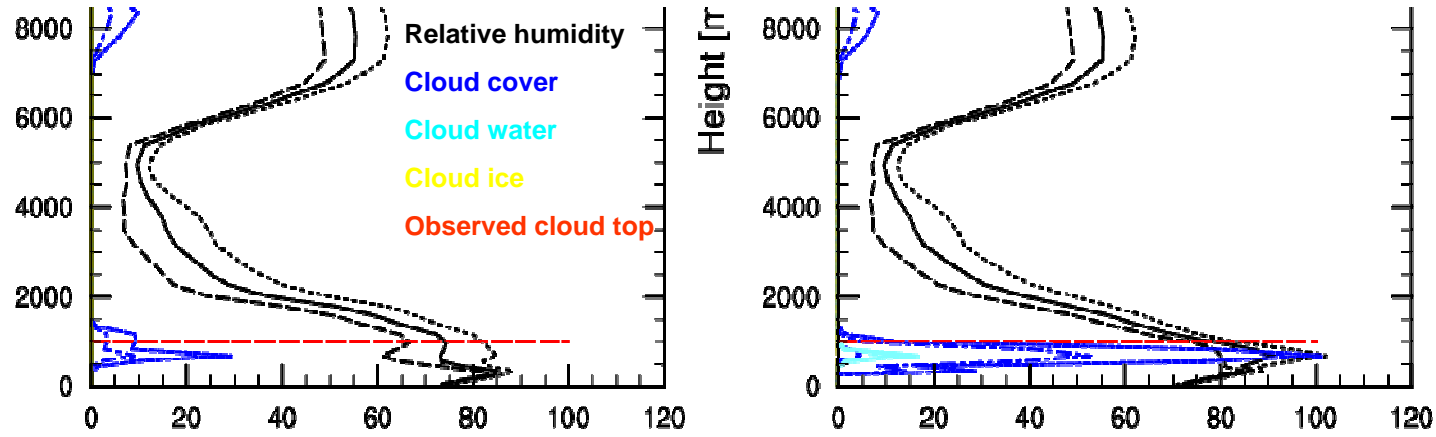


3 lines on one colour indicate mean and mean +/- spread

Increment cross section for the ensemble mean



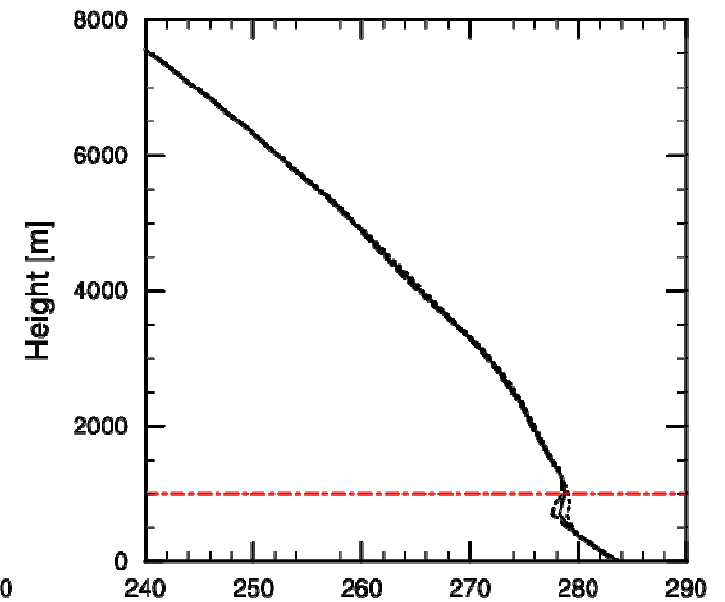
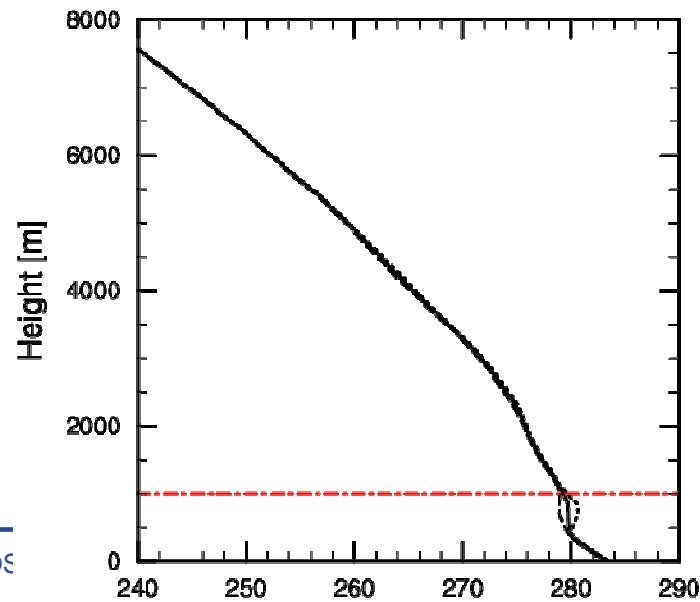
Corresponding temperature Profiles



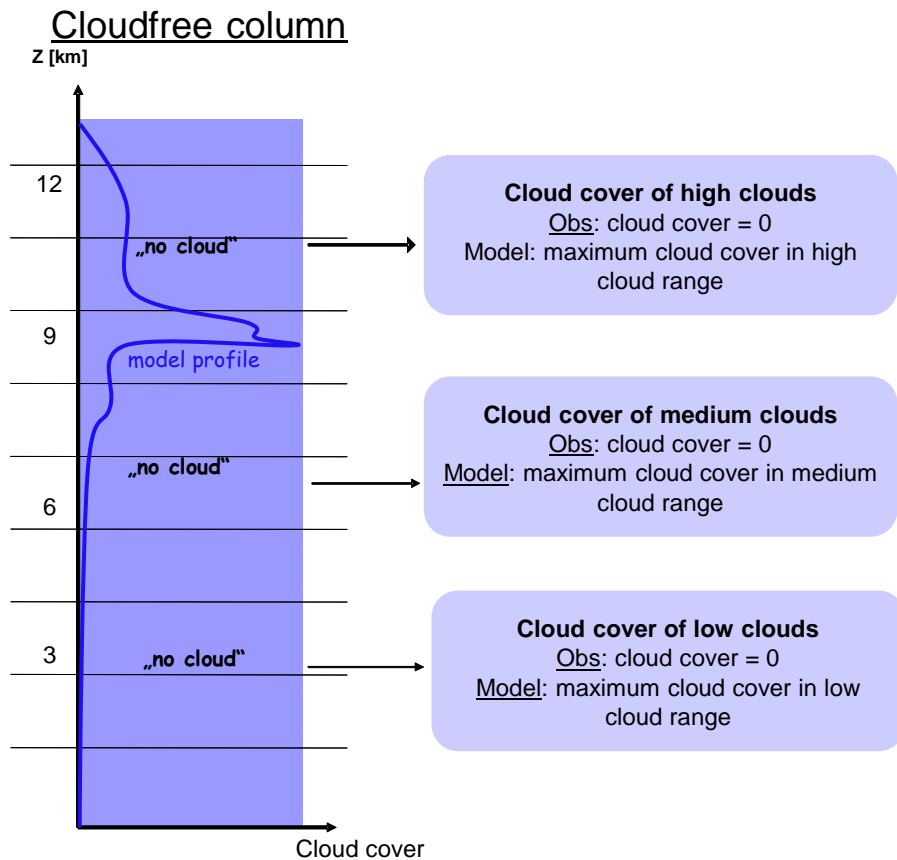
Temperature (mean +/- spread)

FG

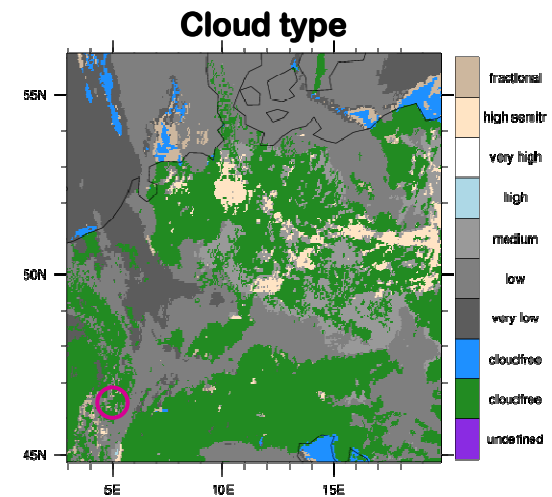
ANA



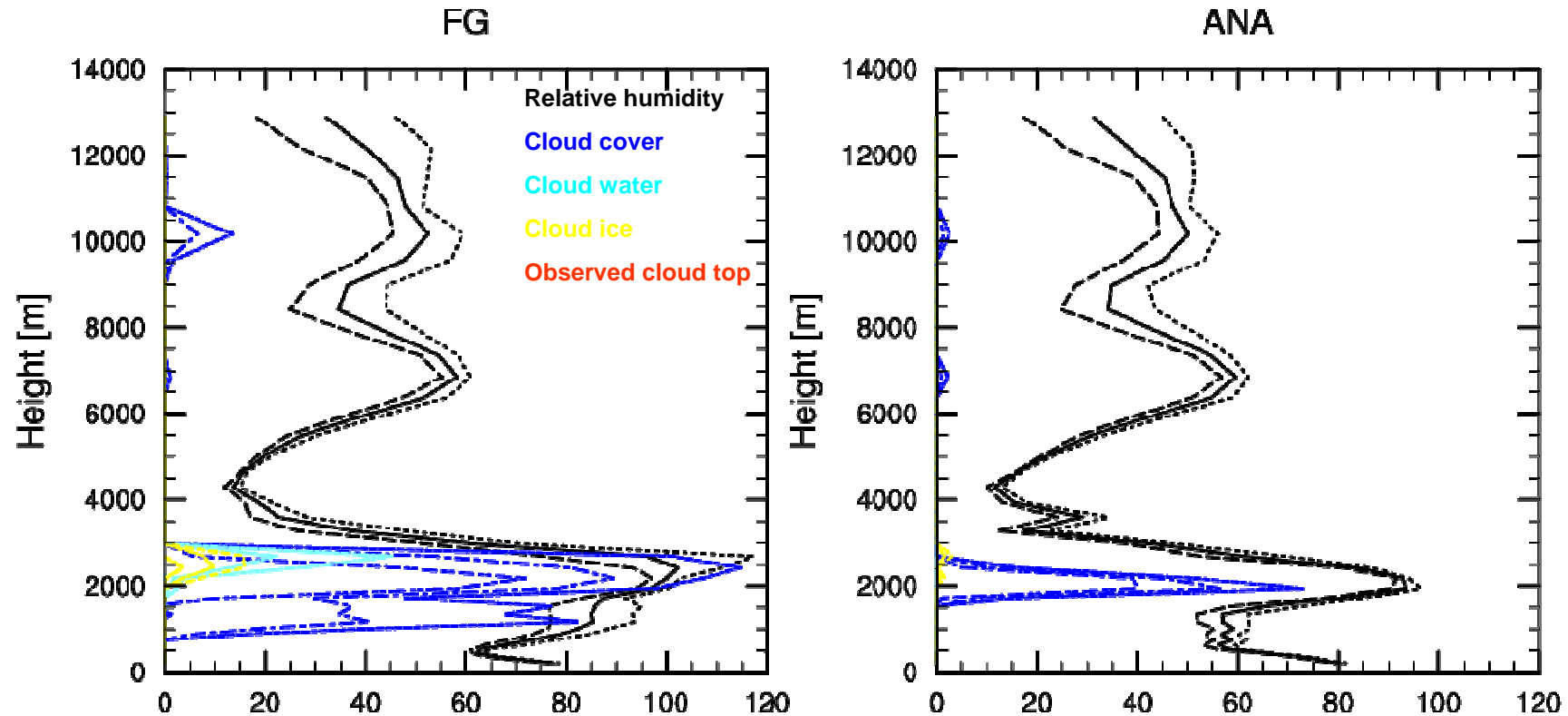
Example 2: „false alarm“ cloud in cloudfree case



4, lon 4.9

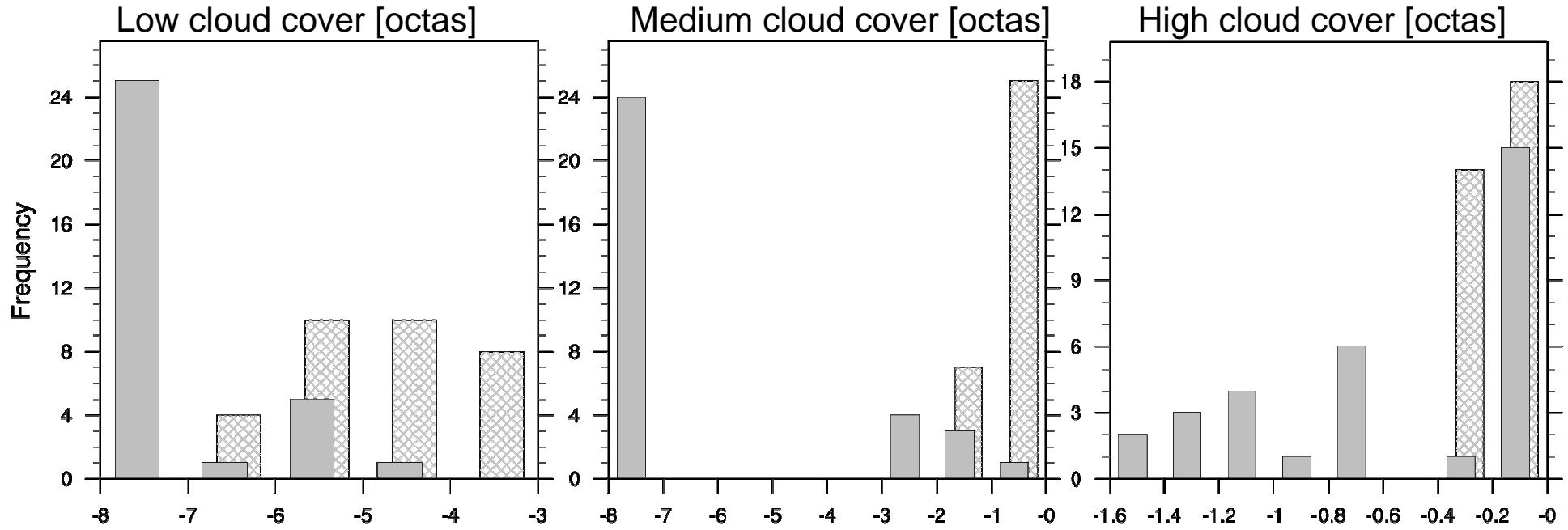


Profiles: mean and spread



3 lines indicate mean and mean +/- spread

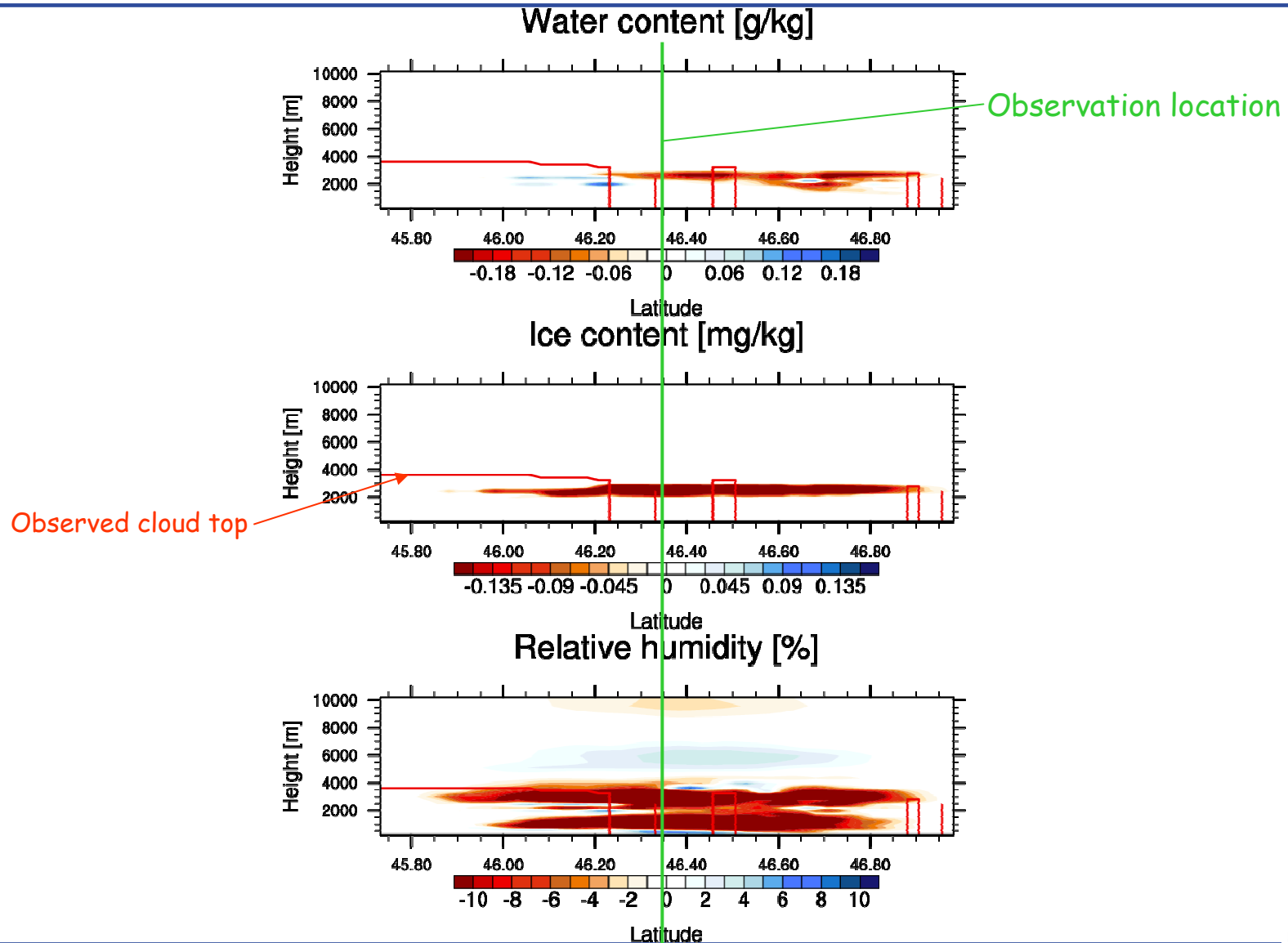
Obs minus model departure histogram



Remember: observed cloud cover = 0



Increment cross section ensemble mean



- **LETKF** offers **new perspectives for assimilating unconventional data**
 - Nudging can only assimilate prognostic model variables, less suitable for indirect observations
 - In the LETKF the **observations are used to weight the different ensemble members** → Assimilate also non-state variables
 - No linearized/adjoint model/physics necessary
 - → Here: assimilation of clouds
 - Can be useful for e.g. stable high-pressure systems in winter, to give higher weights to members which simulate the observed low stratus better
- **Single observation** experiments show:
 - Assimilation of cloud products
 - gives **reasonable, comprehensible results,**
 - **draws ensemble closer to observation**
 - Only small improvement if no ensemble member has cloud (→ physics perturbations needed in ensemble generation...)



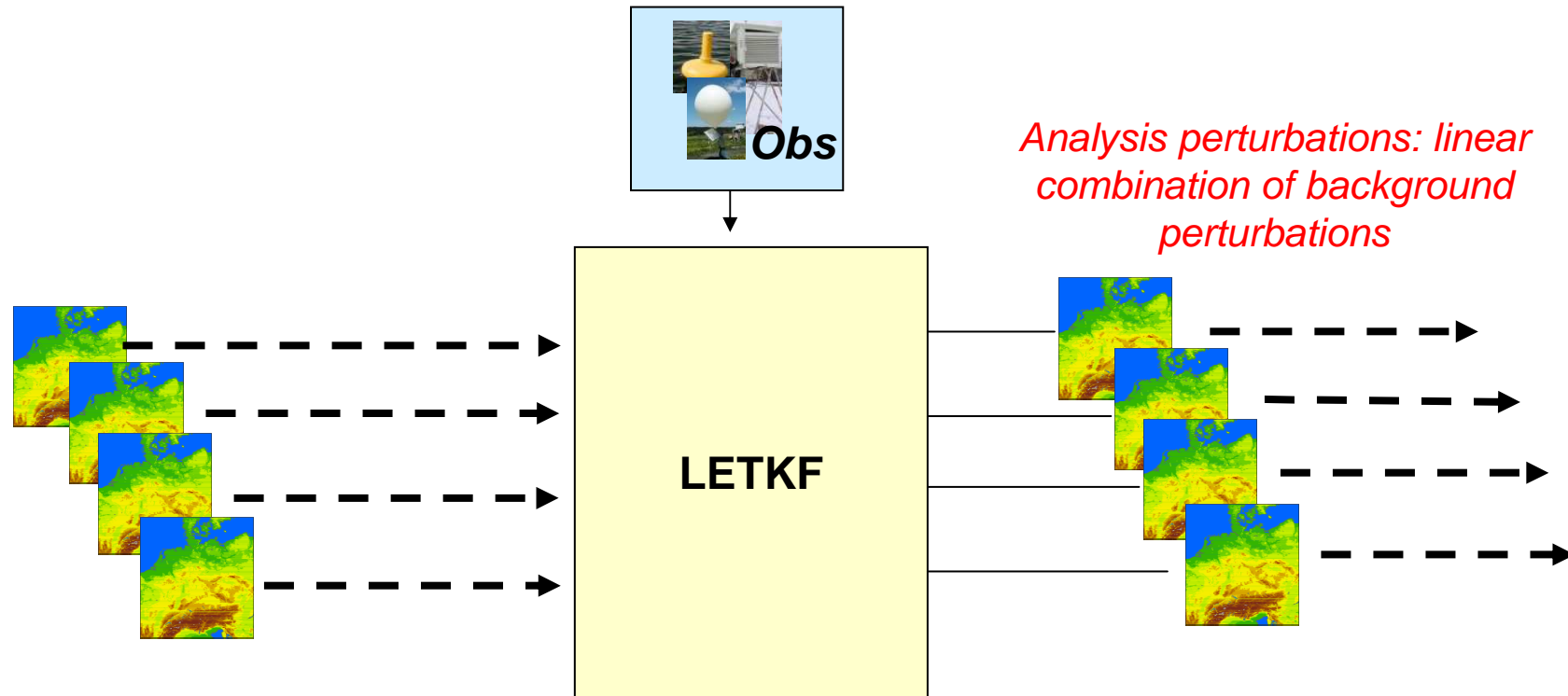
- **Next steps:**
 - Currently: set up of cycled experiments with 40 members
 - With better lateral forcing from GME 40-member LETKF
 - Experiments with dense observations
 - Thinning?
 - Observation errors?
 - Localization?
 - Effects on ensemble spread

Thank you for your attention.





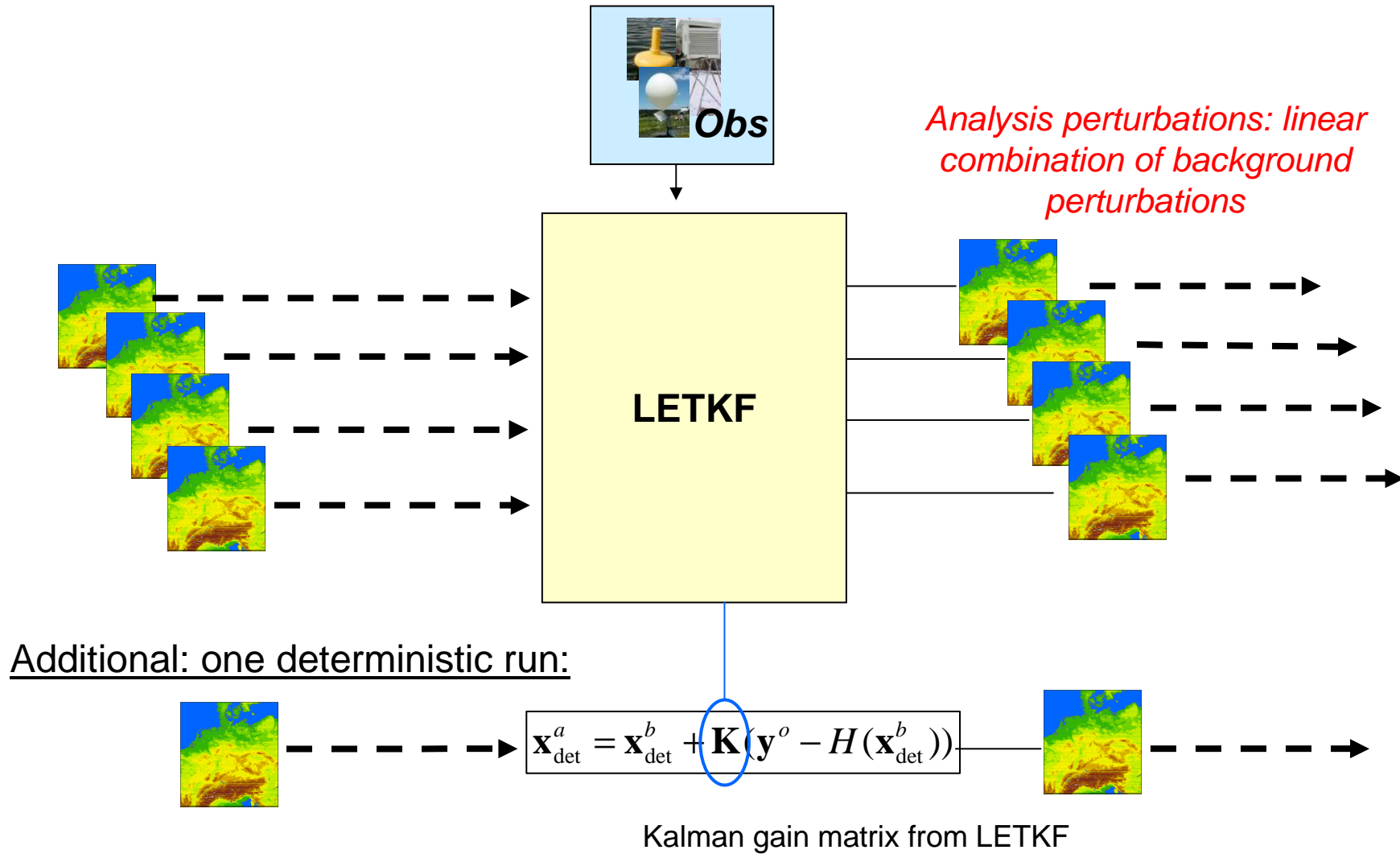
Local Ensemble Transform Kalman Filter in COSMO (KENDA project)



- **L**ocal: the linear combination is fitted in a local region
 - observation have a spatially limited influence region
- **T**ransform: most computations are carried out in ensemble space
 - computational efficient

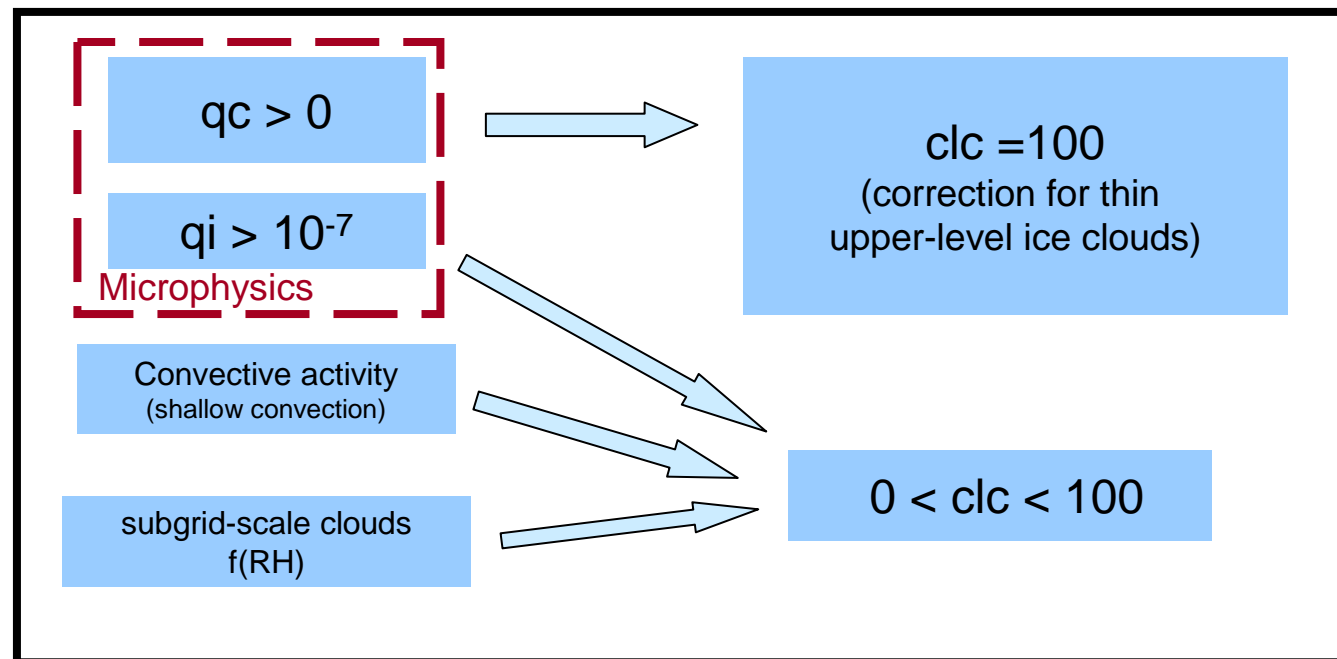


Local Ensemble Transform Kalman Filter



COSMO: cloud parameterization

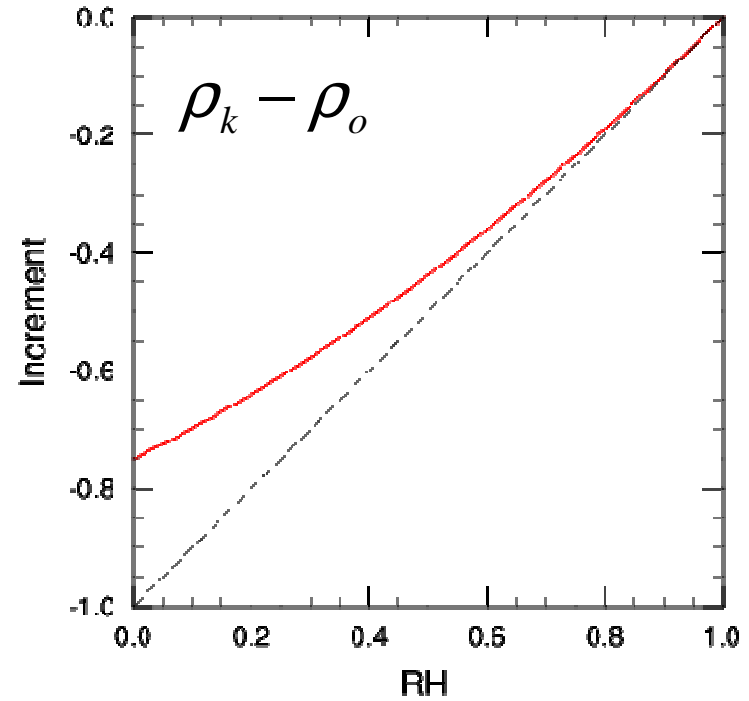
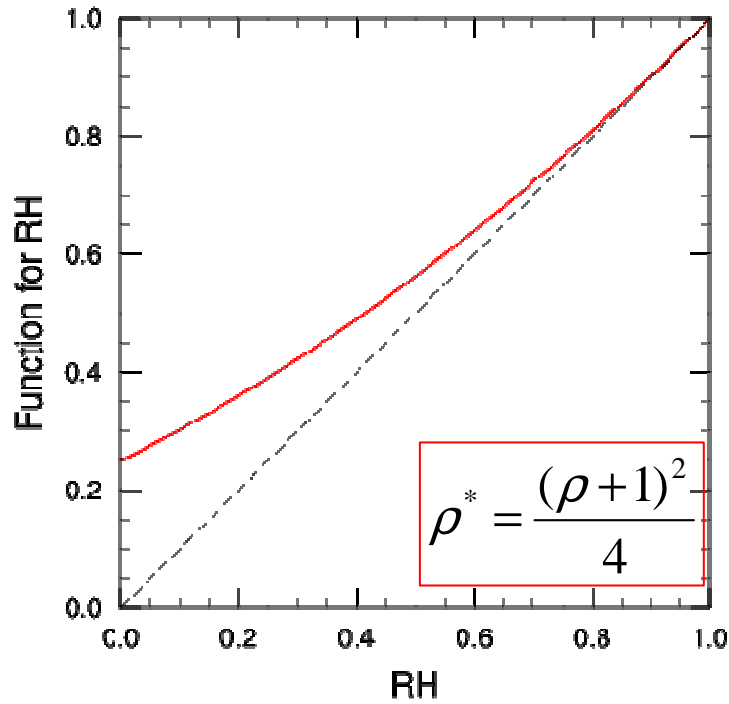
- Relevant variables (3D):
 - Cloud Water q_c [kg/kg]
 - Cloud Ice q_i [kg/kg]
 - Cloud cover cl_c [%]
- } Prognostic model variables



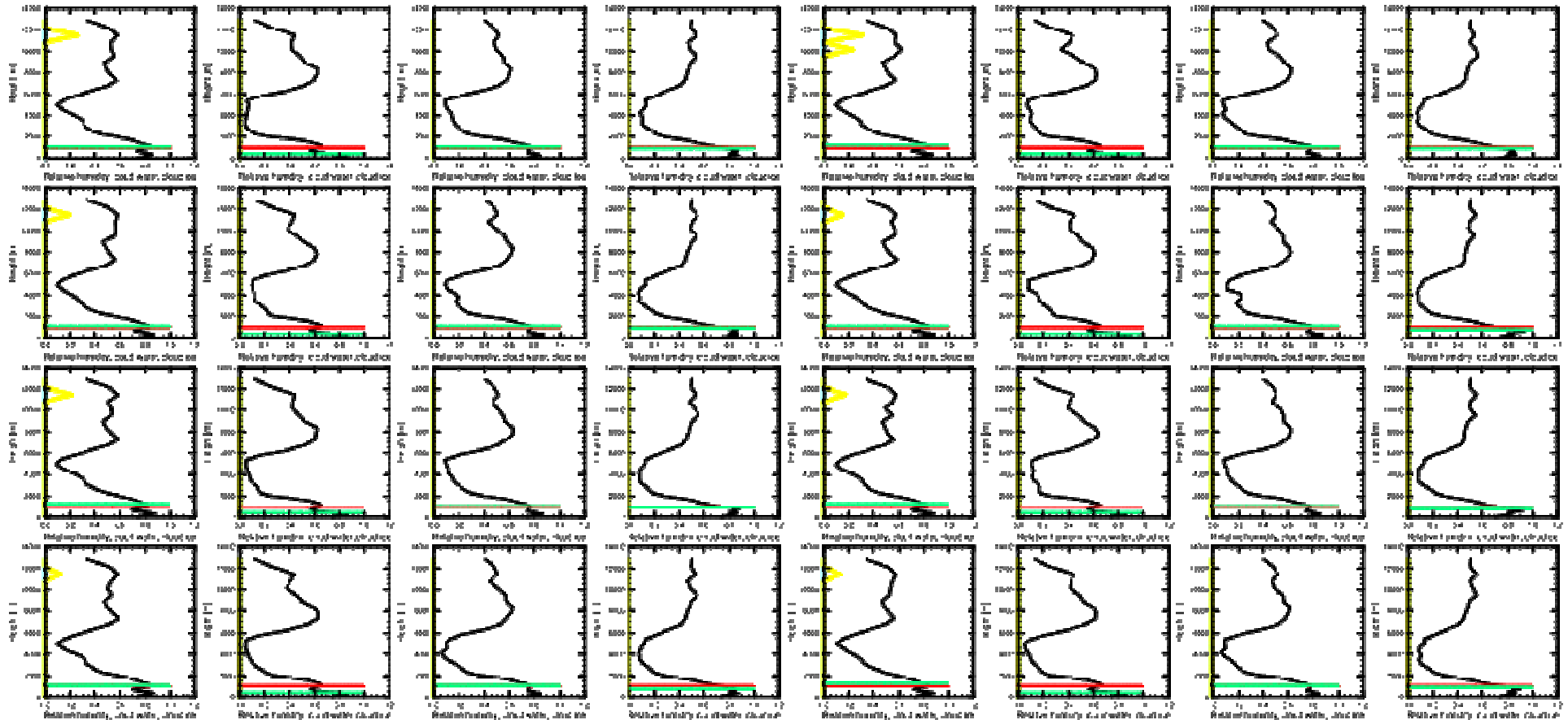
- Weight for cloud top height correction and error correction:

$$\gamma = \frac{(r - r_{max})}{(r + r_{max})} \cdot \frac{(t - t_{max})}{(t + t_{max})}$$

Function for relative humidity

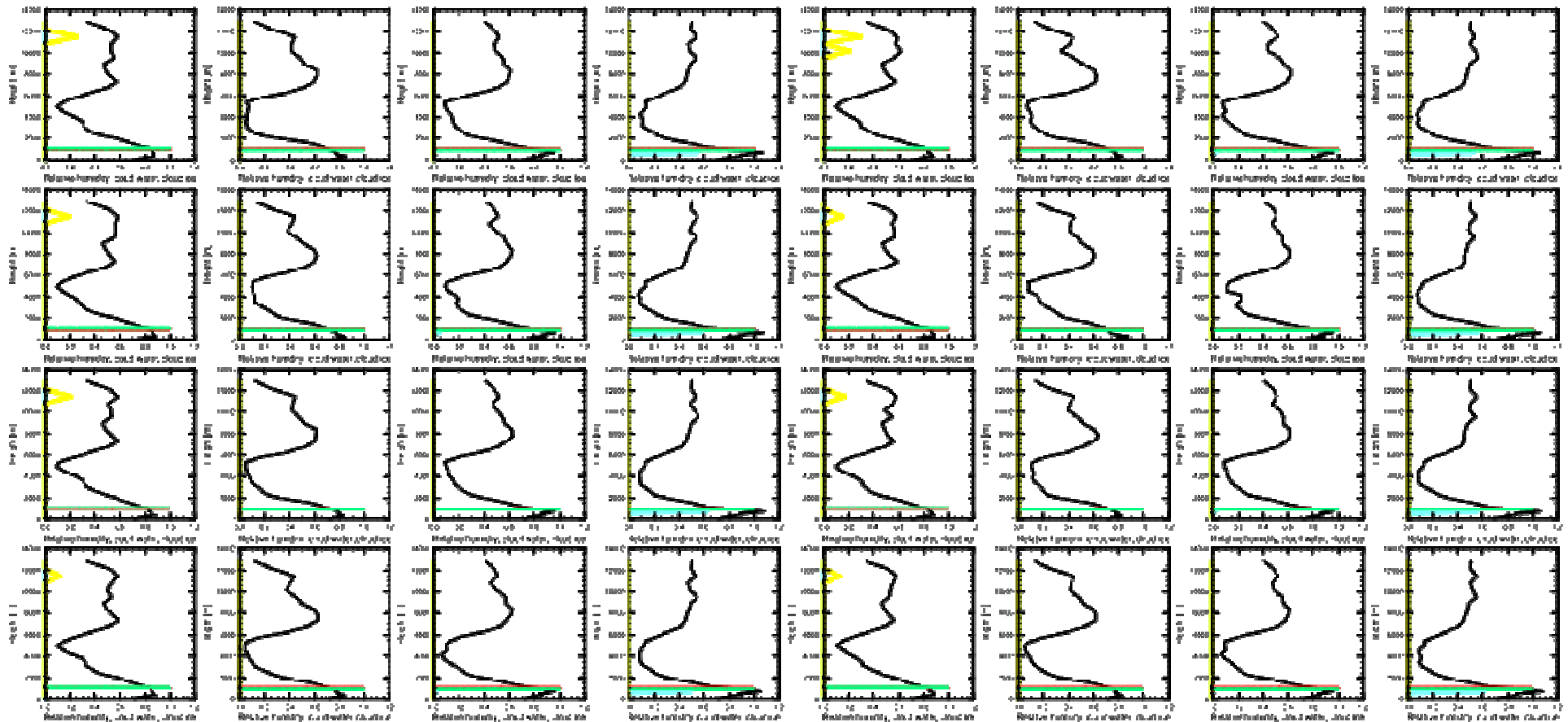


Single profiles: FG



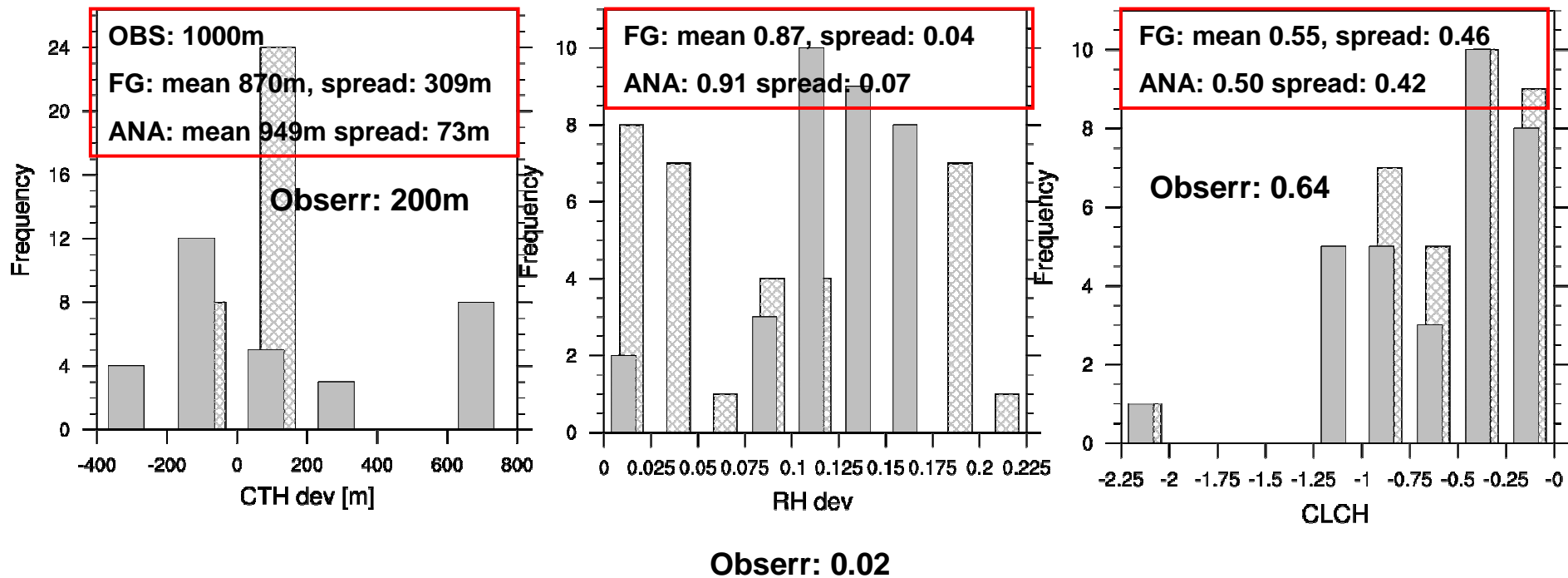
Relative humidity
Cloud water
Cloud ice
Observed cloud top
Model equ. cloud top

Single profiles: ANA



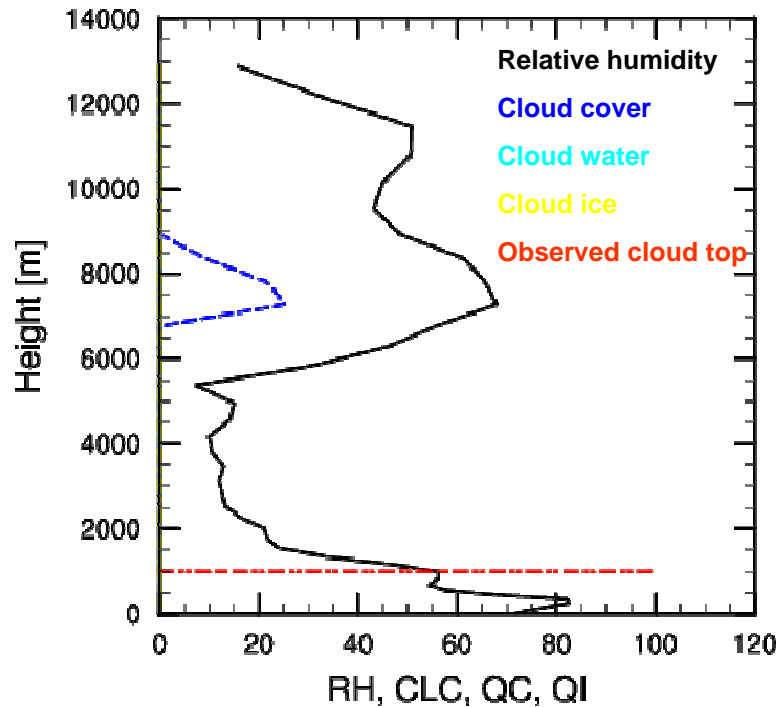
Relative humidity
Cloud water
Cloud ice
Observed cloud top
Model equ. cloud top

Deviation-histograms: Obs minus model, lat 53.9, lon 4.1 (example 1)

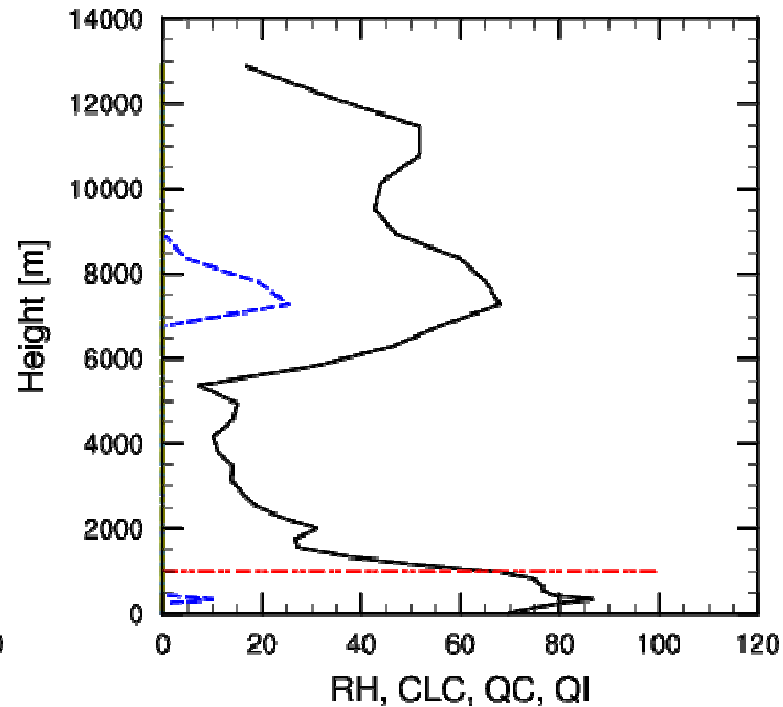


Deterministic run

First guess



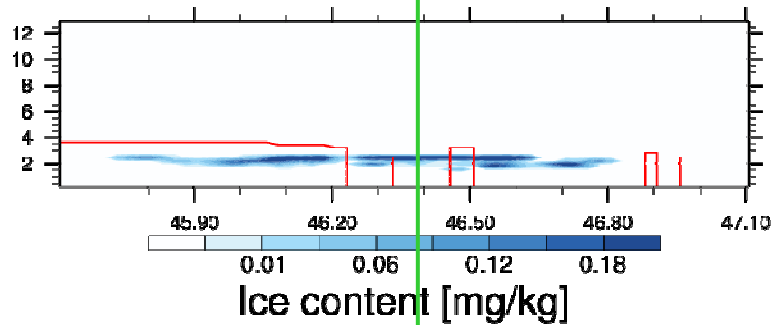
Analysis



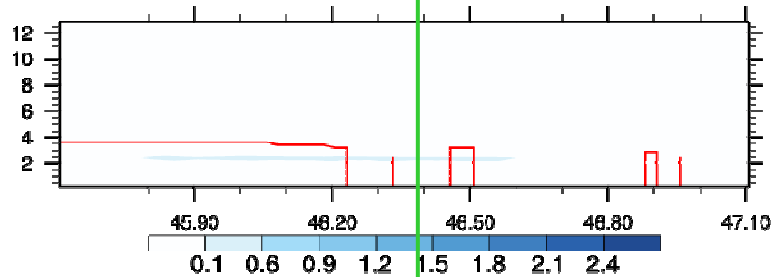
Humidity at cloud layer is increased in deterministic run

Deterministic run, example 2

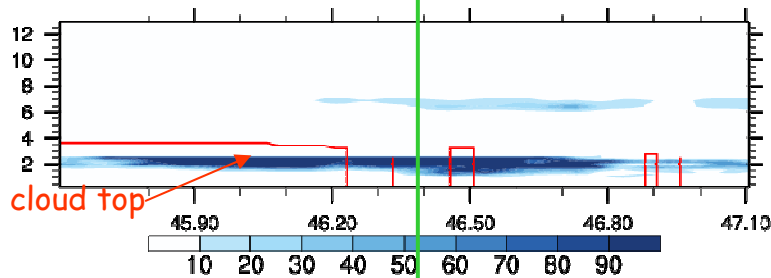
First guess
Water content [g/kg]



Ice content [mg/kg]

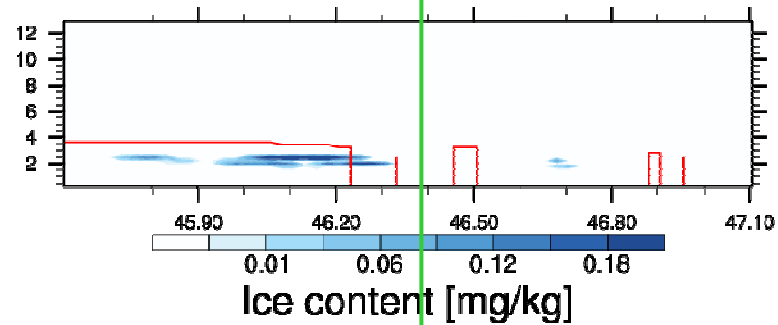


Cloud cover [%]

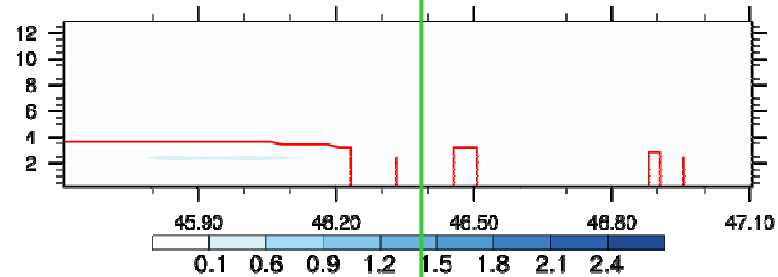


Observed cloud top

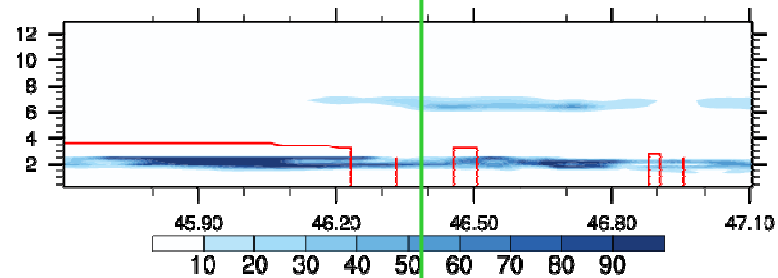
Analysis
Water content [g/kg]



Ice content [mg/kg]

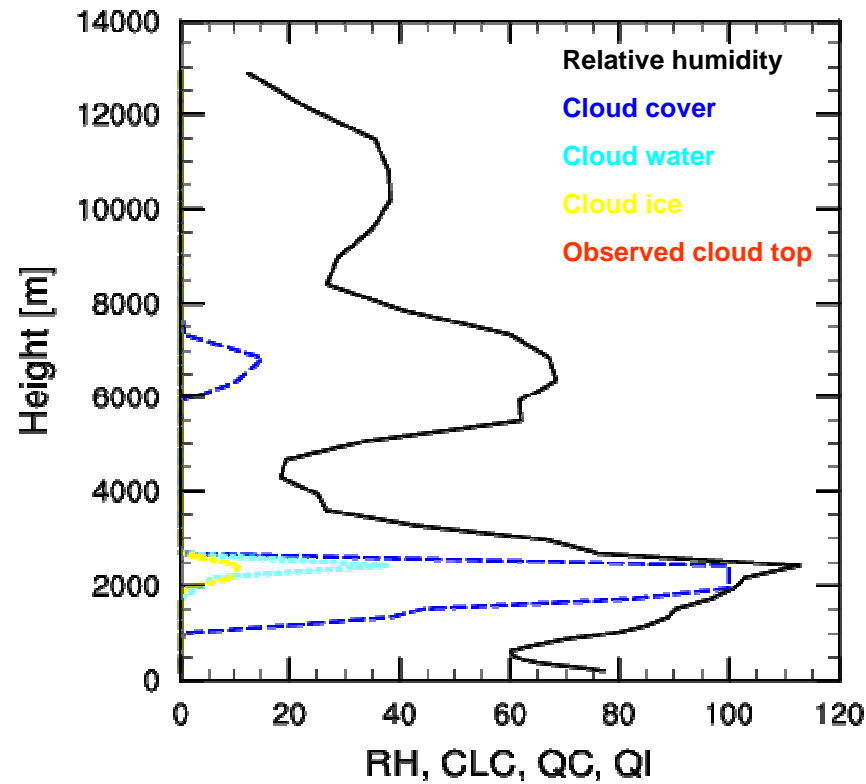


Cloud cover [%]



Deterministic run

First guess



Analysis

