

Characterization of MSG-SEVIRI IR satellite observations and their potential for convective-scale ensemble data assimilation

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Observations for convective-scale data assimilation

- Geostationary satellite observations:
 - high temporal ($\sim$ min) and spatial ($\sim$ km) resolution
 - major source of atmospheric information:
strong sensitivity to temperature, water vapour, clouds
- potential for convective-scale data assimilation

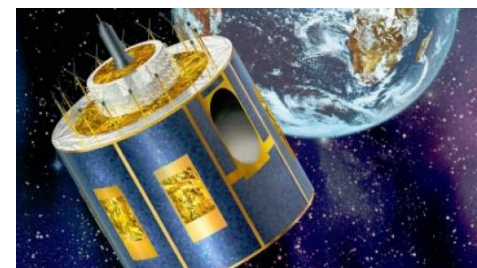
Many challenges for optimal use:

- fast and accurate forward operators
- assimilation of cloud-affected data
(e.g., biases, non-Gaussian errors, difficulties in forward modelling)
- using ensemble data assimilation systems
(e.g., vertical localization, ensemble background errors)

MSG – SEVIRI

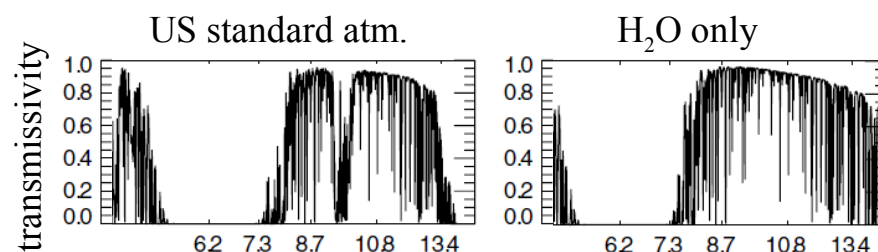
Meteosat Second Generation

Spinning-Enhanced Visible and Infrared Imager

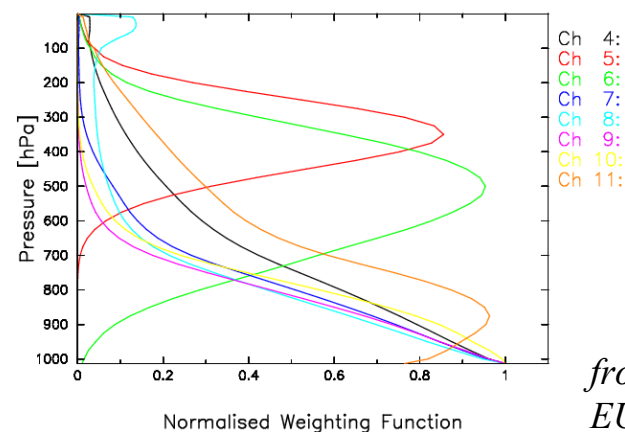


Channel no.		Characteristics of spectral band (μm)			Main gaseous absorber or window
		λ_{cen}	λ_{min}	λ_{max}	
1	VIS0.6	0.635	0.56	0.71	Window
2	VIS0.8	0.81	0.74	0.88	Window
3	NIR1.6	1.64	1.50	1.78	Window
4	IR3.9	3.90	3.48	4.36	Window
5	WV6.2	6.25	5.35	7.15	Water vapor
6	WV7.3	7.35	6.85	7.85	Water vapor
7	IR8.7	8.70	8.30	9.10	Window
8	IR9.7	9.66	9.38	9.94	Ozone
9	IR10.8	10.80	9.80	11.80	Window
10	IR12.0	12.00	11.00	13.00	Window
11	IR13.4	13.40	12.40	14.40	Carbon dioxide
12	HRV	Broadband (about 0.4 – 1.1)			Window/water vapor

from Schmetz et al. 2012



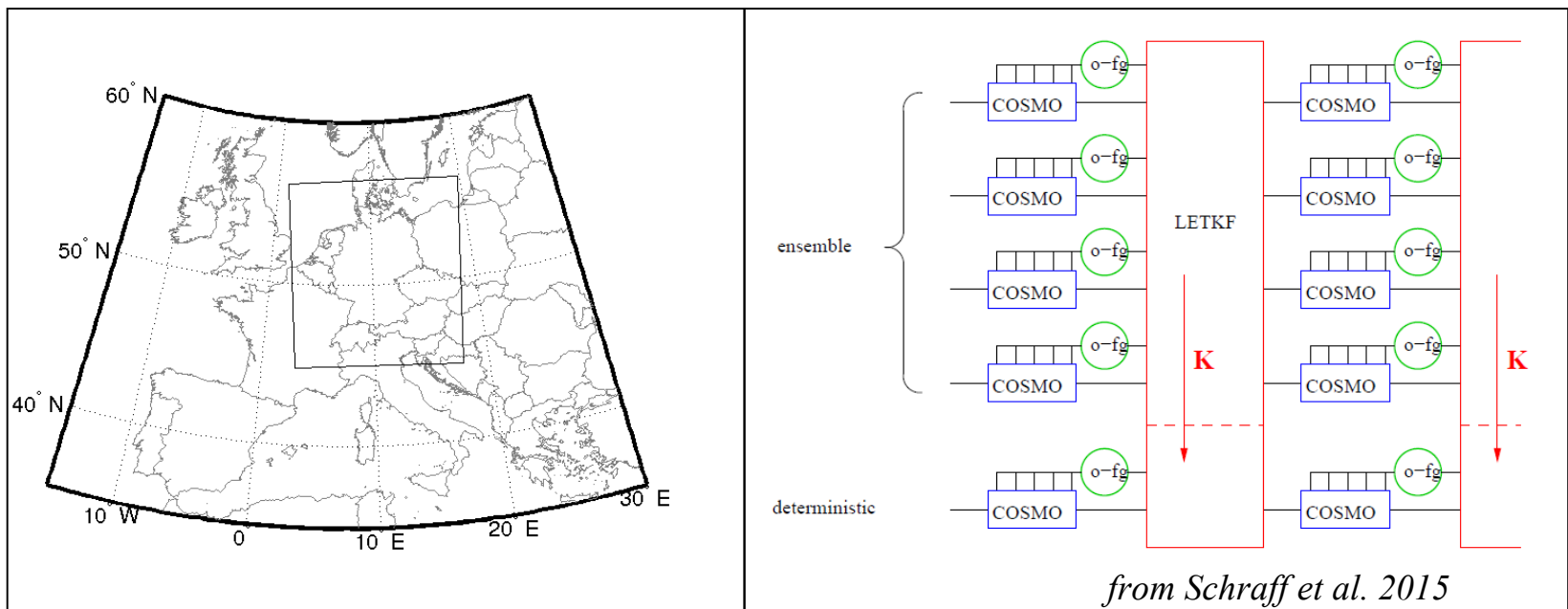
from Stengel 2005



from
EUMETRAIN, 2008

Overall goal:

- Direct assimilation of MSG SEVIRI all-sky observations in the experimental kilometre-scale ensemble data assimilation (KENDA) system of DWD
- Local Ensemble Transform Kalman Filter (LETKF) in the COSMO-DE forecast model at 2.8 km horizontal resolution



Evaluation of satellite observation statistics

Summer time period from 01 – 06 July 2011 over central Europe

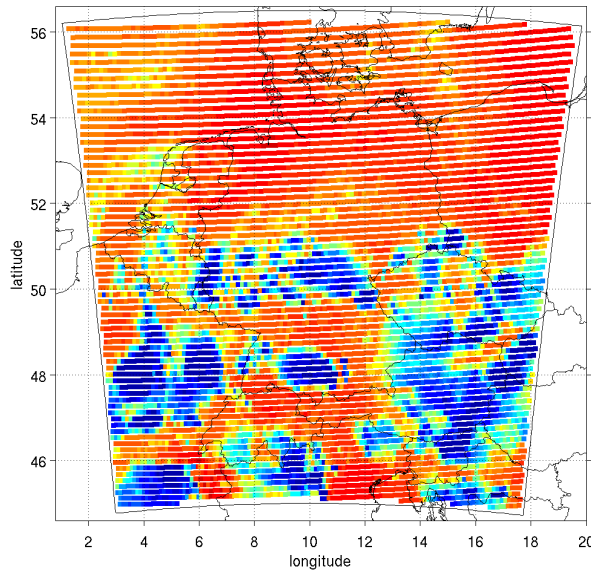
Reference experiment:

- Hourly DA cycle 40 ensemble member LETKF + deterministic run
- Assimilation of conventional observations
- Monitoring of brightness temperatures T_B for SEVIRI channels WV6.2, **WV7.3** at reduced horizontal resolution of ~ 10 km
- Radiative transfer with RTTOV 10:
 - (i) multiple scattering scheme for cloudy scenes: fractional cloud cover, cloud liquid and ice water content (LWC, IWC)
 - (ii) Wyser (1998) parametrisation to convert IWC to effective diameter
→ potentially larger errors / biases in cloudy situations

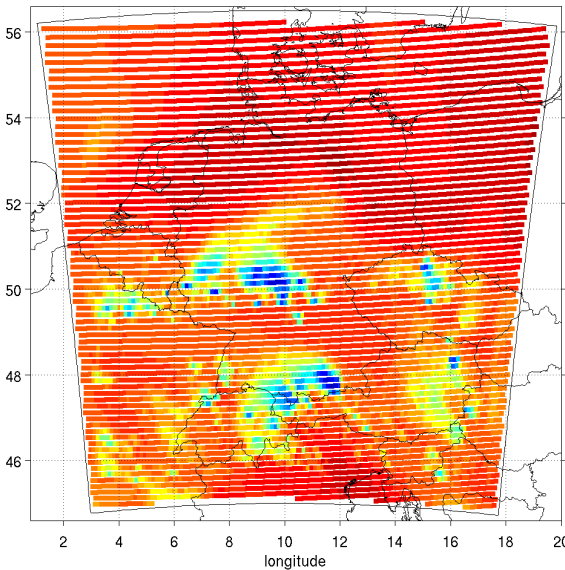
Example case 16 UTC 04 June 2011

WV7.3 T_B (K):

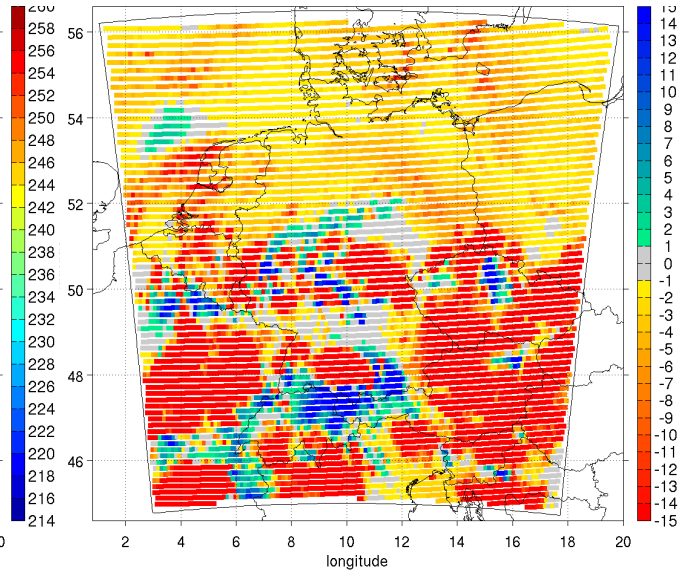
observation



model



observation - model

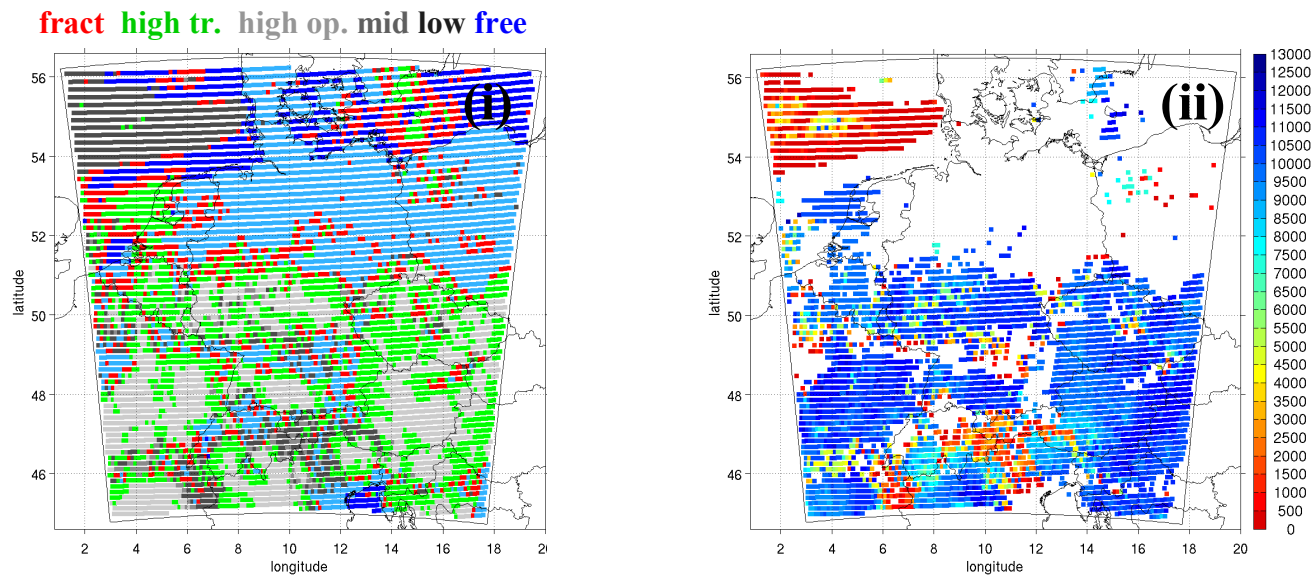


- Significant differences between observed and simulated T_B
- Errors seem to differ for clear and cloudy areas

→ Evaluate performance and correct potential bias of simulated T_B

Evaluation of satellite observation statistics

- Nowcasting Satellite Application Facility (NWC SAF):
(i) cloud type and (ii) **cloud top height** product for MSG-SEVIRI data



- Model cloud top height** estimation from cloud LWC / IWC:
model level where cloud water / ice content $> 1 \text{ mg m}^{-3}$

Evaluation of satellite observation statistics

- Static bias correction derived from monitoring statistics as function of cloud type:

y^{BC} for T_B (K) at WV 7.3 μm :

-2.9 K **clear**

-2.9 K **low (< 2 km AGL)**

-3.1 K **mid (2-6 km AGL)**

-4.4 K **high (> 6 km AGL)**

from Perianez 2014

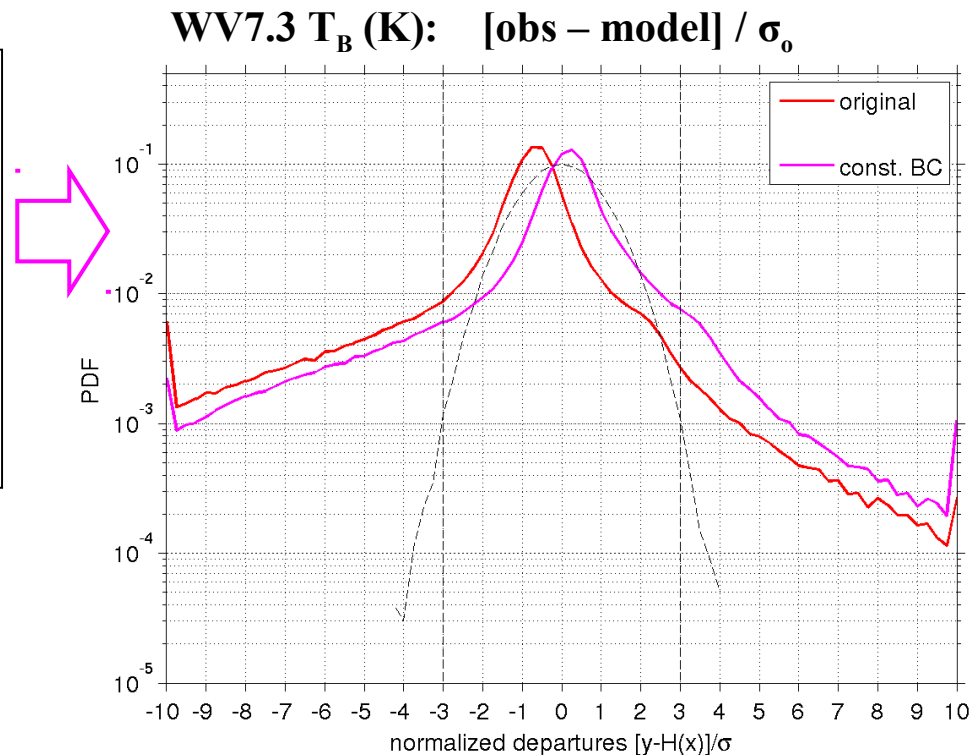
Evaluation of satellite observation statistics

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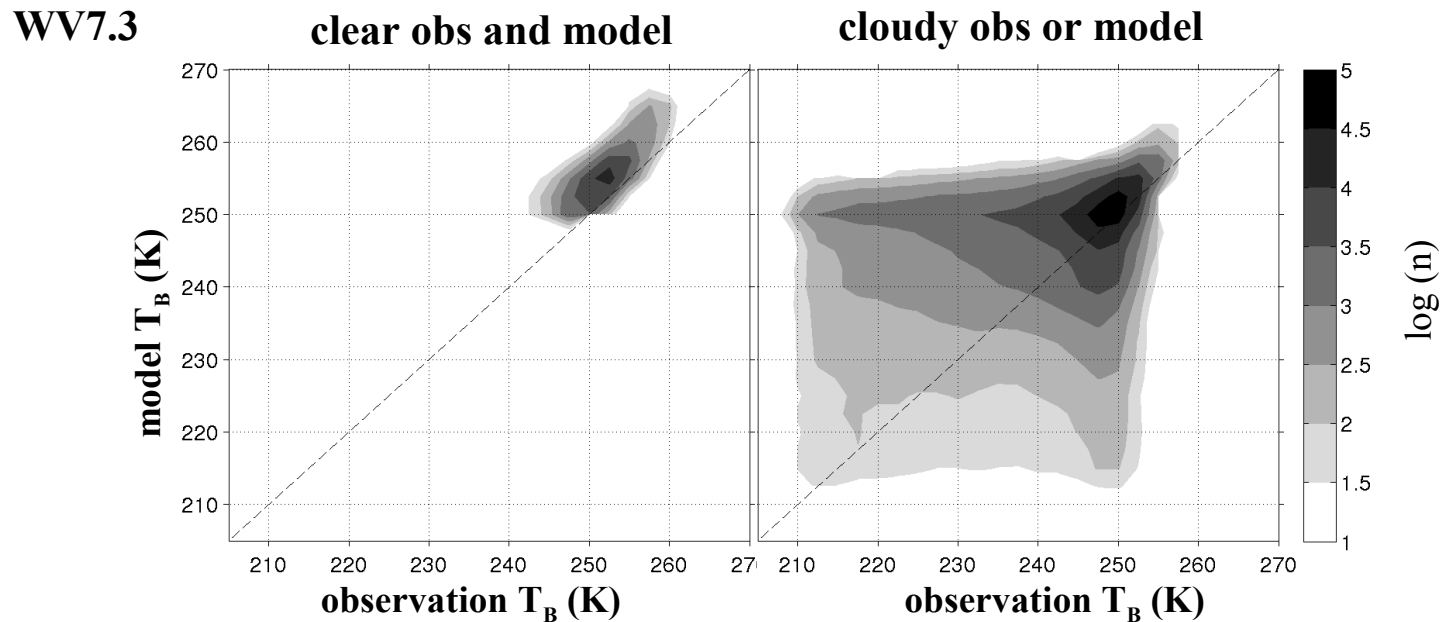
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from Perianez 2014



→ Current bias is corrected, but actual values are likely to change

Error model for all-sky data assimilation

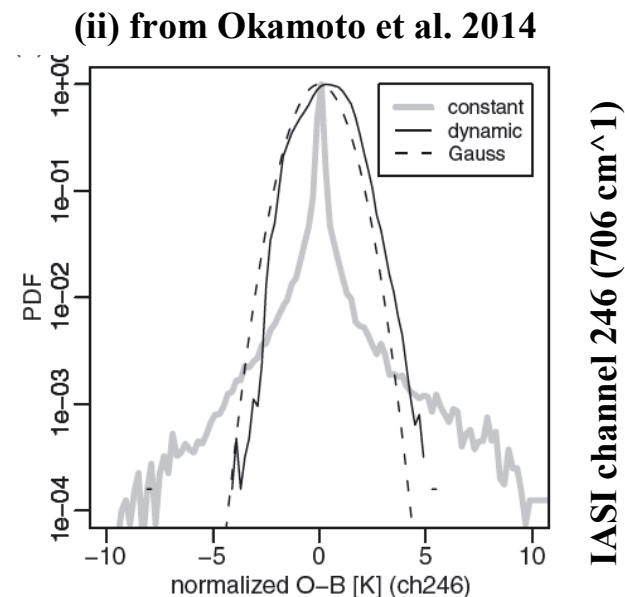
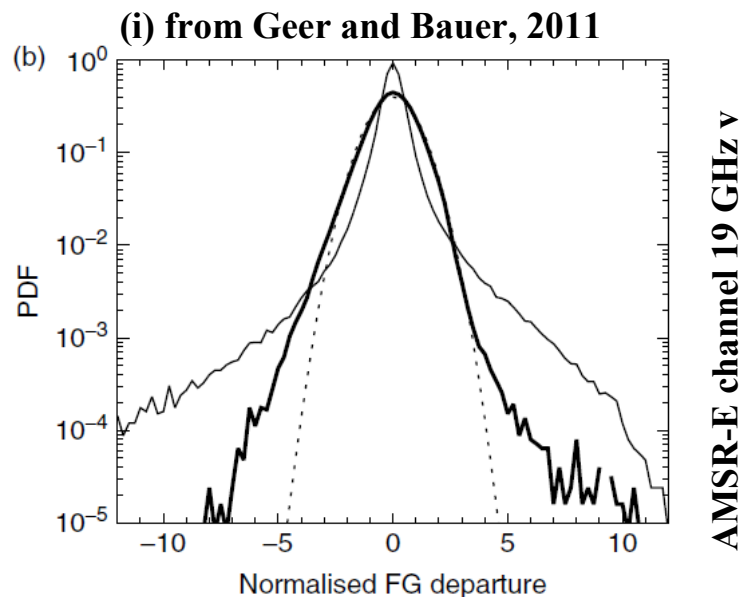


- Agreement between obs and model depend on the cloud type / height
- Forecasts are not very accurate in predicting exact cloud type / height at the correct location

Error model for all-sky data assimilation

Can we predict how big the cloud impact is on the first-guess departures?

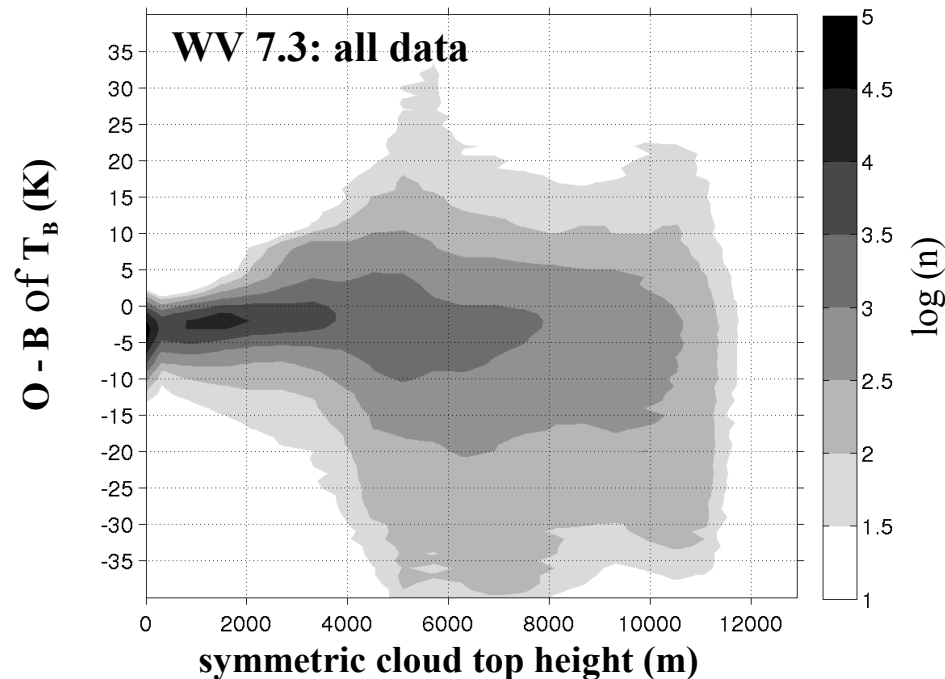
- Different predictors used before for observation errors:
 - (i) symmetric cloud amount or (ii) cloud effect average
- used for quality control (QC) and observation error assignment



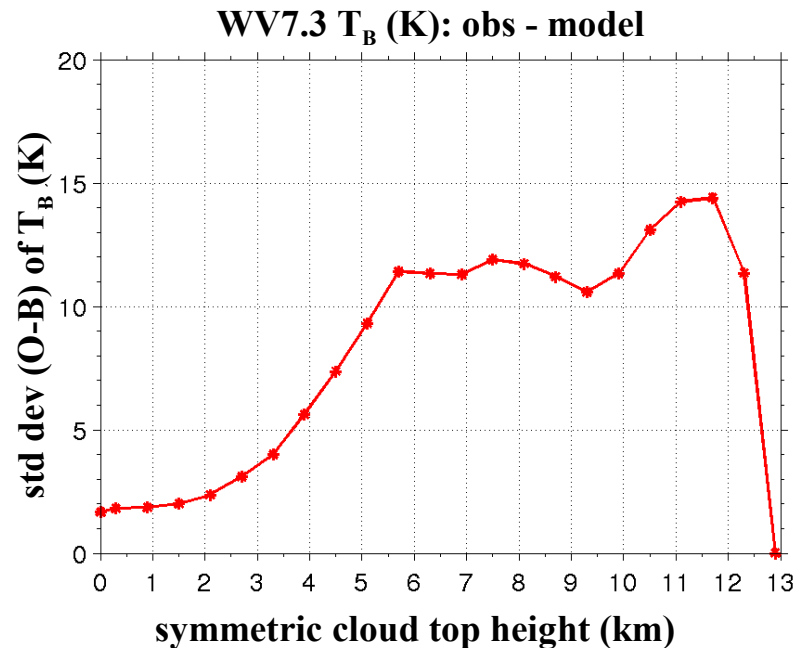
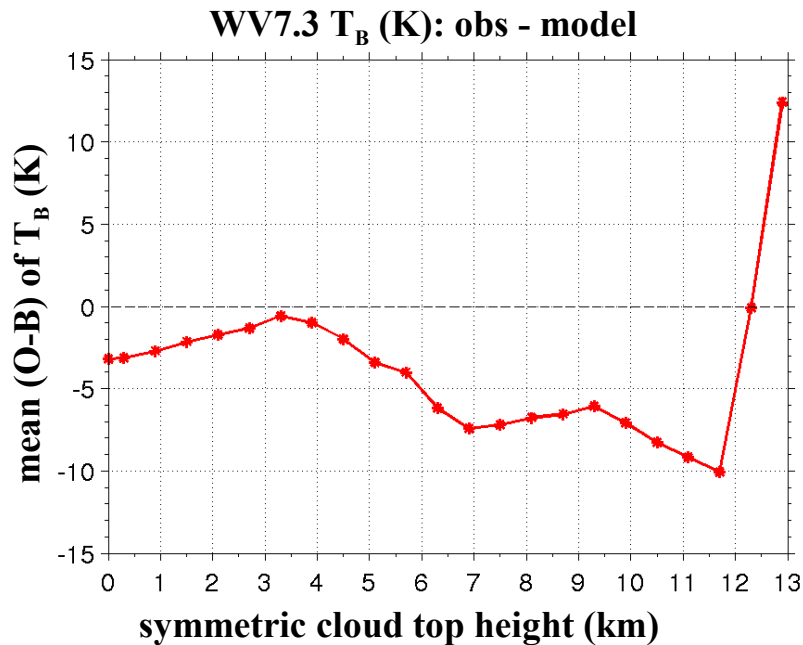
Error model for all-sky data assimilation

- Use cloud top height information from the observation and the model
- Define symmetric cloud top height

$$\text{cth_sym} = (\text{cth_obs} + \text{cth_model}) / 2$$

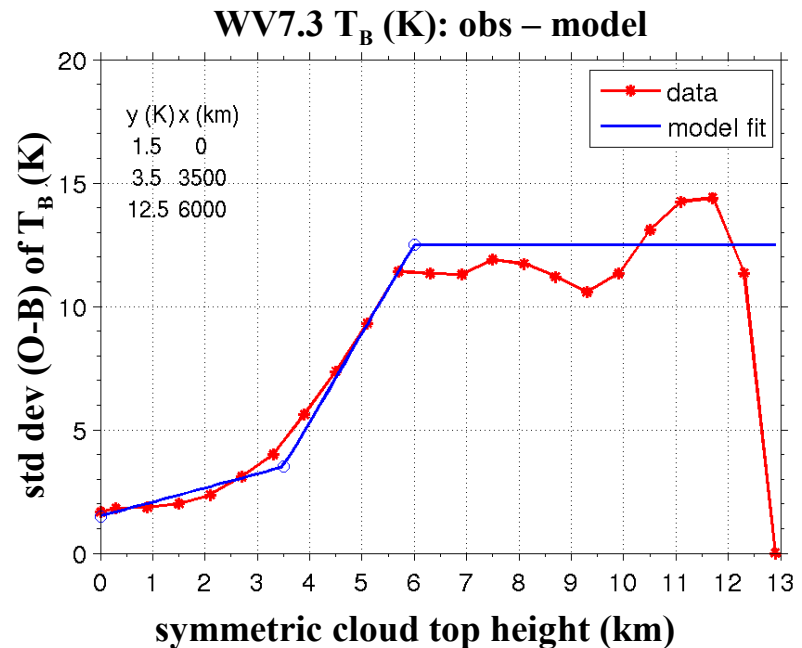
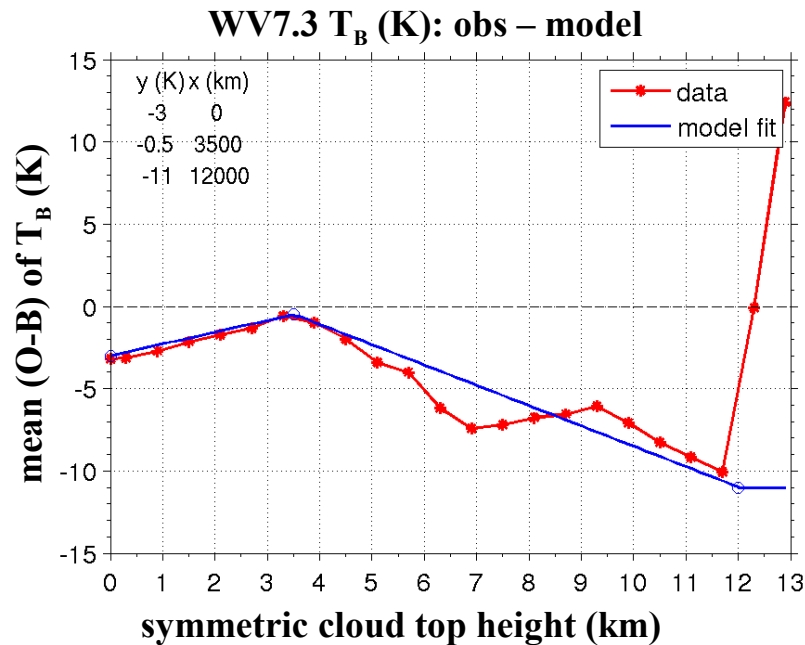


Error model for all-sky data assimilation



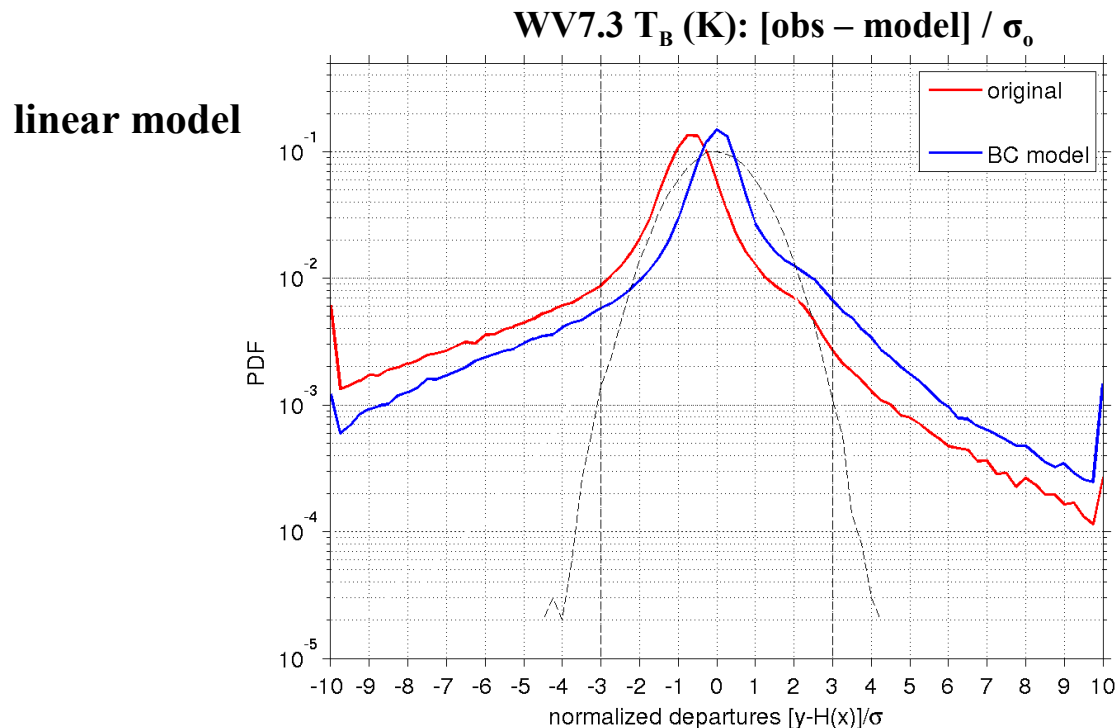
- Compute mean and standard deviation for different cth_sym bins
→ cth_sym can predict bias and standard deviation of O-B departures

Error model for all-sky data assimilation



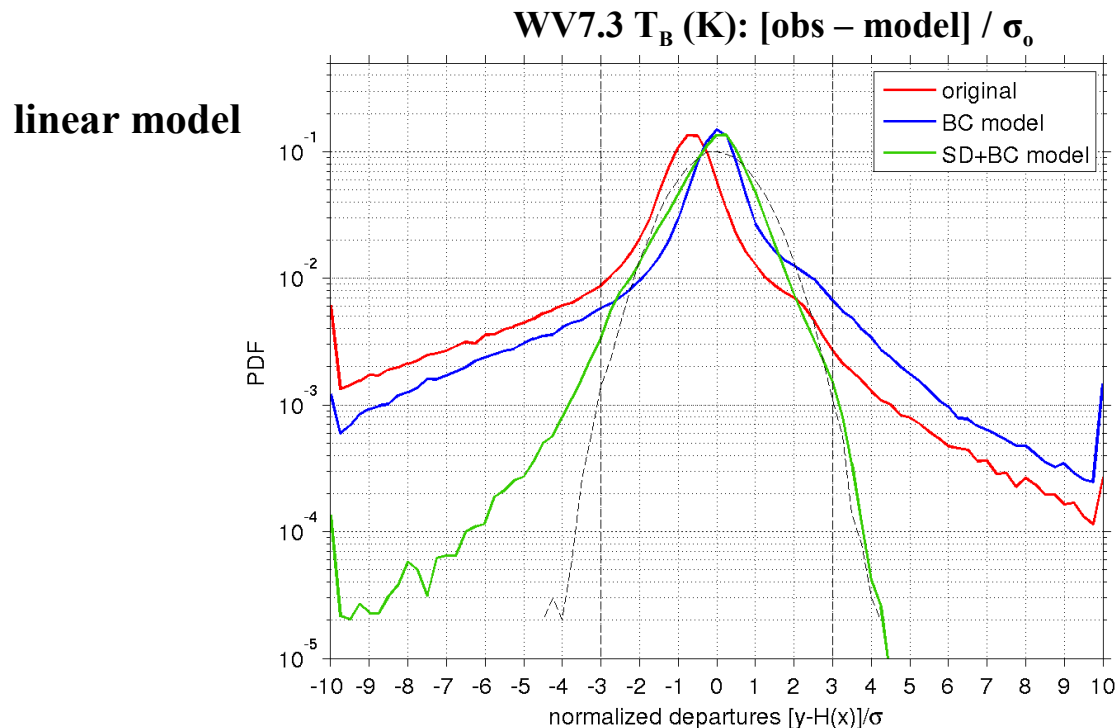
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→ cth_sym can predict bias and standard deviation of O-B departures
- **Linear fit** to model standard deviation and bias as function of cth_sym

Error model for all-sky data assimilation



- **Uncorrected O-B departures** are non-Gaussian and biased
- **Bias correction (BC)** as function of `cth_sym` reduces the bias

Error model for all-sky data assimilation

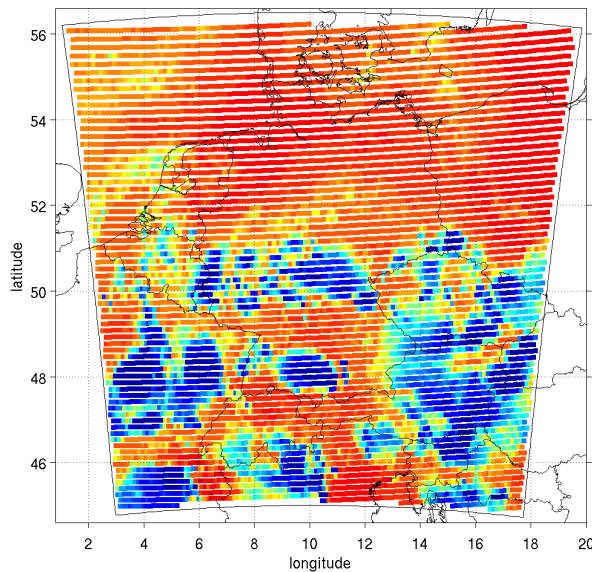


- **Uncorrected O-B departures** are non-Gaussian and biased
- **Bias correction (BC)** as function of cth_sym reduces the bias
- **Dynamic observation error** produces more Gaussian O-B departures

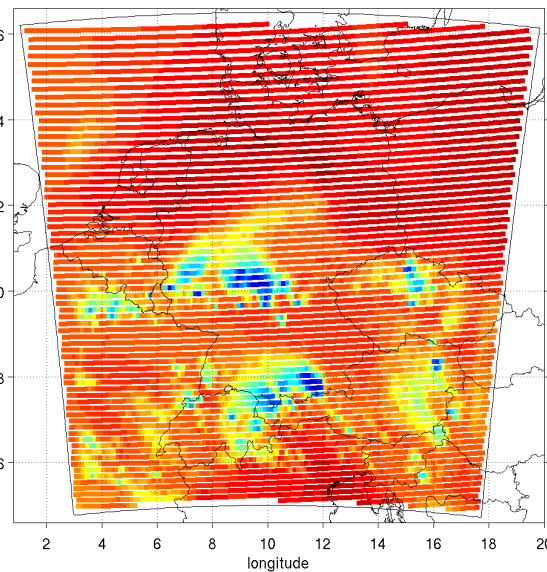
Bias correction on 16 UTC 04 June 2011

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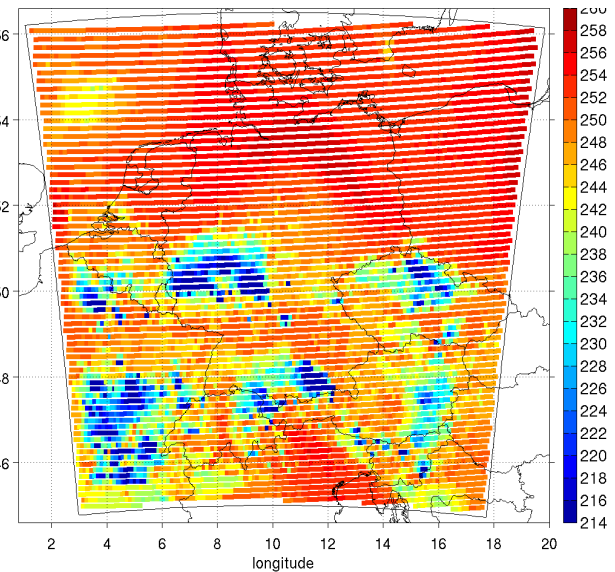
observation



model



bias corrected model



- Model background seems to produce too few high-level clouds
- Overestimation of brightness temperatures, model is too warm
- BC model tries to reduce the warm model bias (clear sky & cloudy)

Bias correction of T_b in ensemble data assimilation

- Biases depend on the presence of a cloud and the cloud type
→ cloud dependent bias correction needed
- Bias correction is usually applied to the observations: $y + y_{bc}$
- Different ensemble members might forecast different cloud situation
→ apply the bias correction to the model instead of the observation

$$H(\mathbf{x}_b^l) + y_{bc}^l$$

→

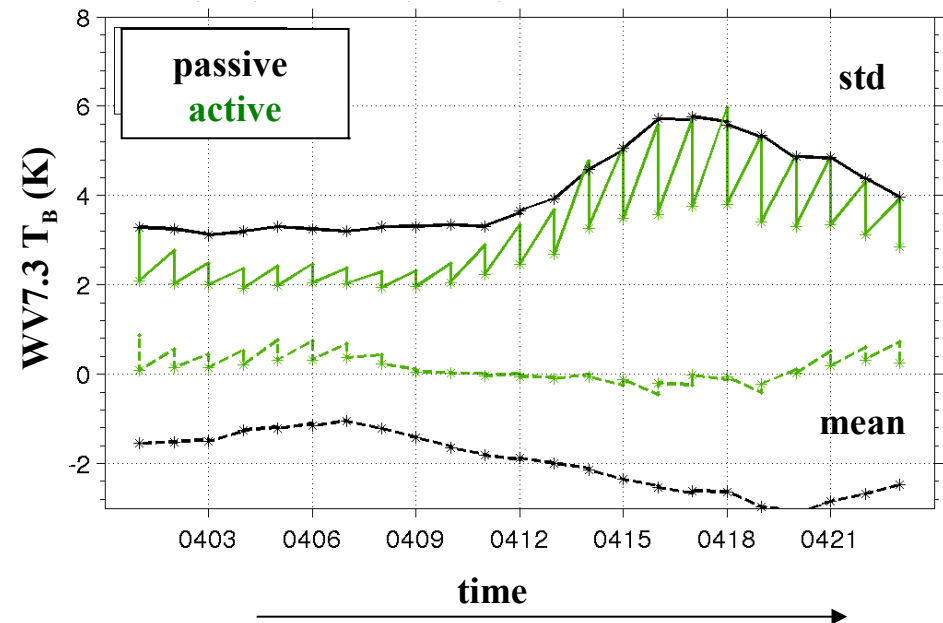
$$\bar{\mathbf{x}}_a = \bar{\mathbf{x}}_b + \mathbf{K} [\mathbf{y} - H(\bar{\mathbf{x}}_b)]$$

$$\bar{\mathbf{x}}_a = \bar{\mathbf{x}}_b + \mathbf{K} \left[\mathbf{y} - \frac{1}{L} \sum_{l=1}^L (H(\mathbf{x}_b^l) + y_{bc}^l) \right]$$

$$\bar{\mathbf{x}}_a = \bar{\mathbf{x}}_b + \mathbf{K} \left[\mathbf{y} - H(\bar{\mathbf{x}}_b) - \frac{1}{L} \sum_{l=1}^L y_{bc}^l \right]$$

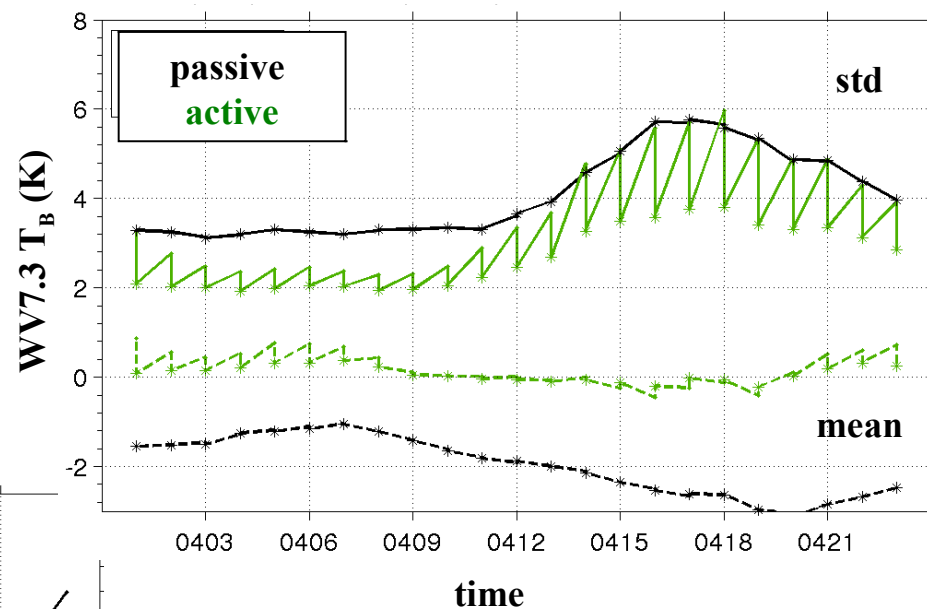
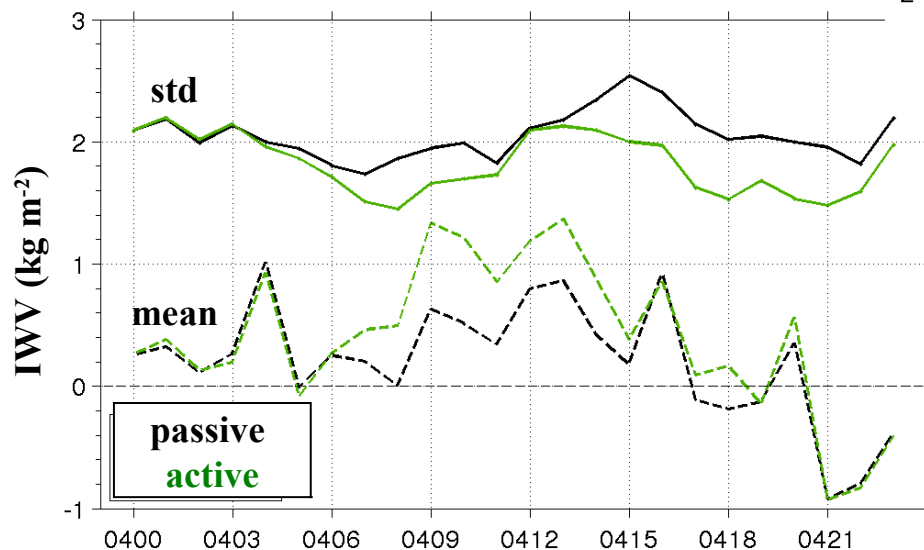
Preliminary results of data assimilation

conv. obs.	active, as in reference
SEVIRI T_B	WV7.3 active ($\Delta x \sim 10$ km)
bias correction	cth_sym (km) = [0,3.5,12] y^{bc} (K) = [-3,-0.5,-11]
obs. error	3.5 K
localization	$l_h = 35$ km, $l_v = 0.7 \log p$



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1-h FC of
Integrated Water Vapour
verified against
GNSS surface stations

Summary

- Direct assimilation of all-sky MSG-SEVIRI brightness temperatures requires careful pre-processing before assimilation
 - large cloud dependent biases and observation errors
- Error model for assimilation of all-sky observations: use `cth_sym` to predict the expected observation statistics
 - bias / observation error can be modelled as function of `cth_sym`
 - cloud dependent bias correction shows good results
 - predicted observation error estimate can be used in the QC and data assimilation and should lead to more Gaussian statistics
- Preliminary results show that all-sky data assimilation of WV7.3 channel works in principle ...

Outlook

- Include the **observation error estimate** in the QC and data assimilation
→ several observation error estimates possible in the ensemble
- **Single-observation experiments**: investigate the behaviour of the LETKF for different szenarios (e.g, cloudy in obs, clear in ensemble)
- Alternative predictors for the observation error model
→ different way to determine cth in model and obs, not optimal
- **Super-obbing** of SEVIRI observations → try to reduce non-linearities
- Sensitivity of simulated COSMO-DE T_B to changes in the parametrization (i) of microphysics and (ii) subgrid scale effects
→ impact on cloud properties, cloud cover, LWC/IWC, ...
- Compare findings between **VIS and WV (IR) channels** for same period

References

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