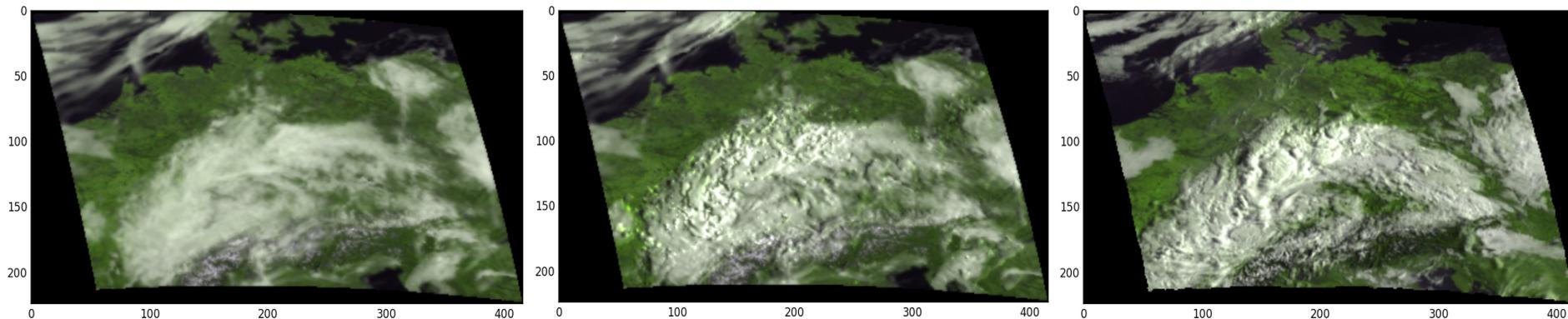


# Using visible and near-infrared satellite observations for convective-scale data assimilation

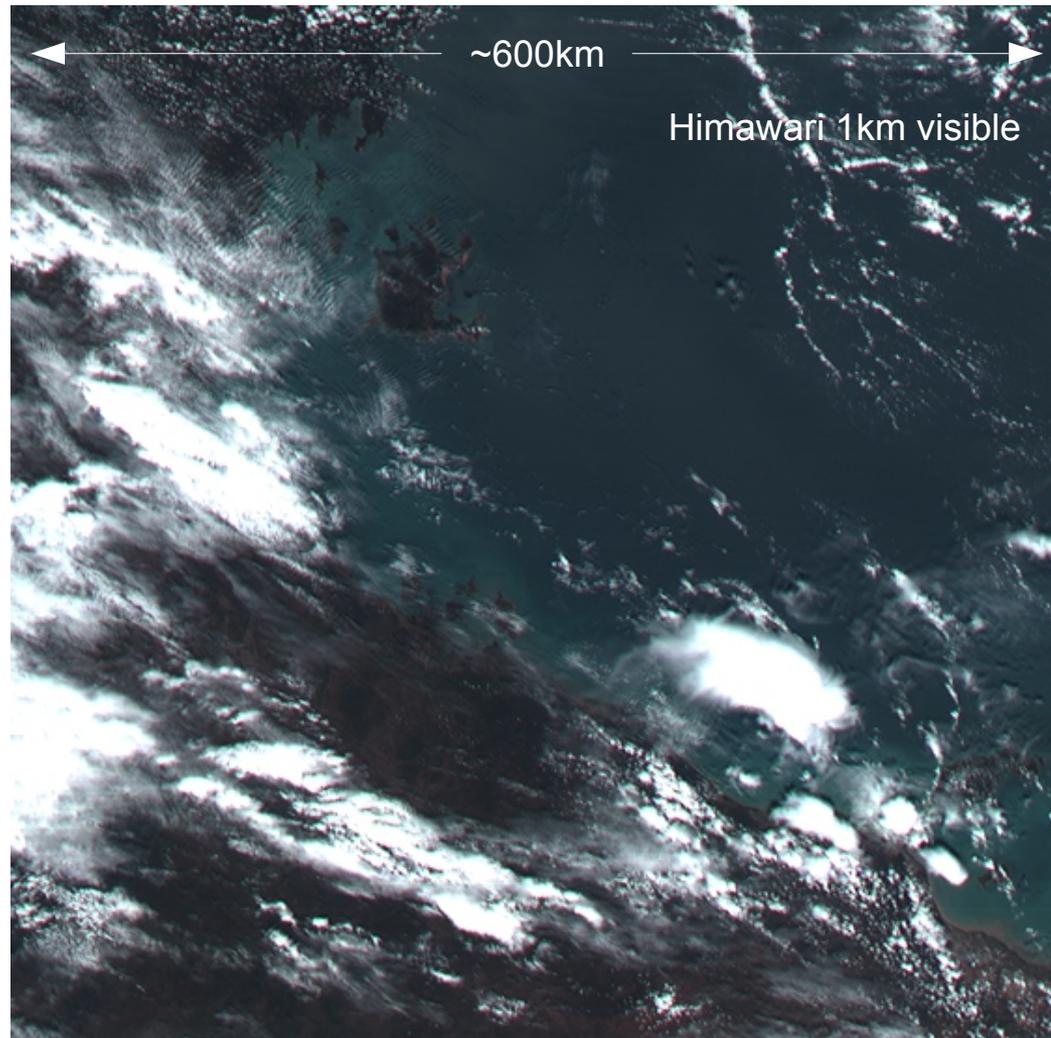
Leonhard Scheck<sup>1,2</sup>, Bernhard Mayer<sup>2</sup>, Martin Weissmann<sup>1,2</sup>

- 1) Hans-Ertel-Center for Weather Research, Data Assimilation Branch
- 2) Ludwig-Maximilians-Universität, Munich



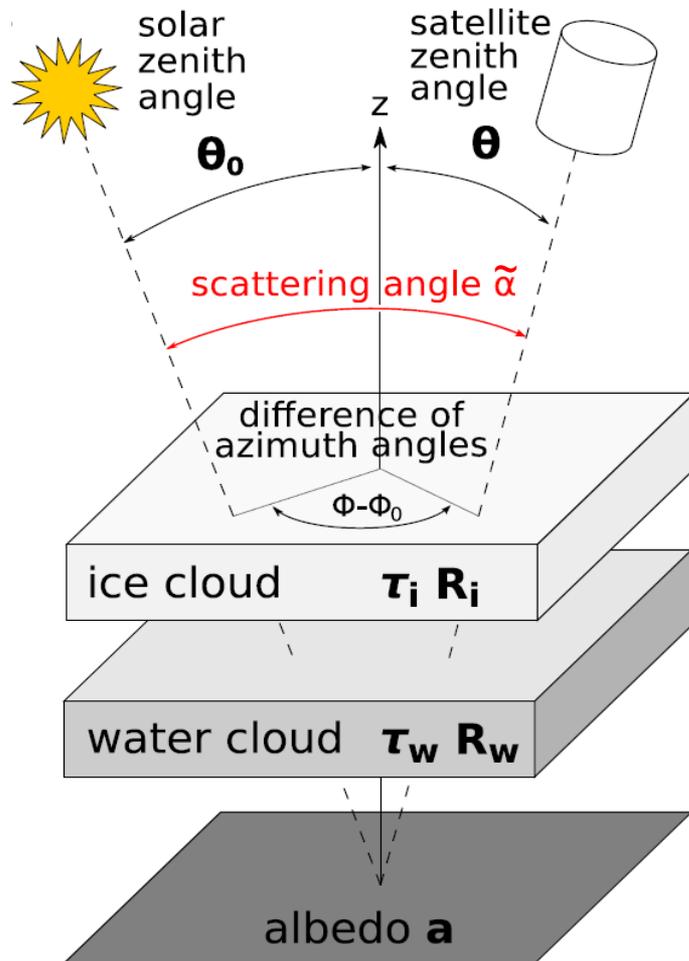
## Visible / near-infrared satellite observations for DA

- relevant for **convective scale DA**:  
high spatial and temporal resolution.  
Himawari-8/9, GOES-R, MTG:  
0.6 $\mu\text{m}$  resolution: 500m (IR: 2km)  
6-8 of 16 channels  $\lambda < 4\mu\text{m}$   
full disc in 5min, target area 30sec
- provide complementary information on **cloud distribution** (convection earlier visible than in radar, low clouds clearly detectable), **cloud properties** (particle size, water phase) and **cloud structure**
- Solar channels are not assimilated in operational DA: **fast forward operators not available** (scattering makes radiative transfer complex)  
→ operator development at HErZ



# Strategy for fast radiative transfer method MFASIS

Method for Fast  
Satellite Image  
Synthesis



## Simplifications

### - Simplified Equation:

3D RT  $\rightarrow$  1D RT (plane-parallel, independent columns)

Computational effort for one Meteosat SEVIRI image:

CPU-days (3D Monte Carlo)  $\rightarrow$  CPU-hours (1D DISORT)

### - Simplified vertical structure:

Cloud water and ice can be separated to form two homogeneous clouds at fixed heights without changing reflectance significantly

$\rightarrow$  only 4 parameters (optical depth, particle size)

+ 3 angles, albedo  $\rightarrow$  **8 parameters per column**

## Reduction of computational effort

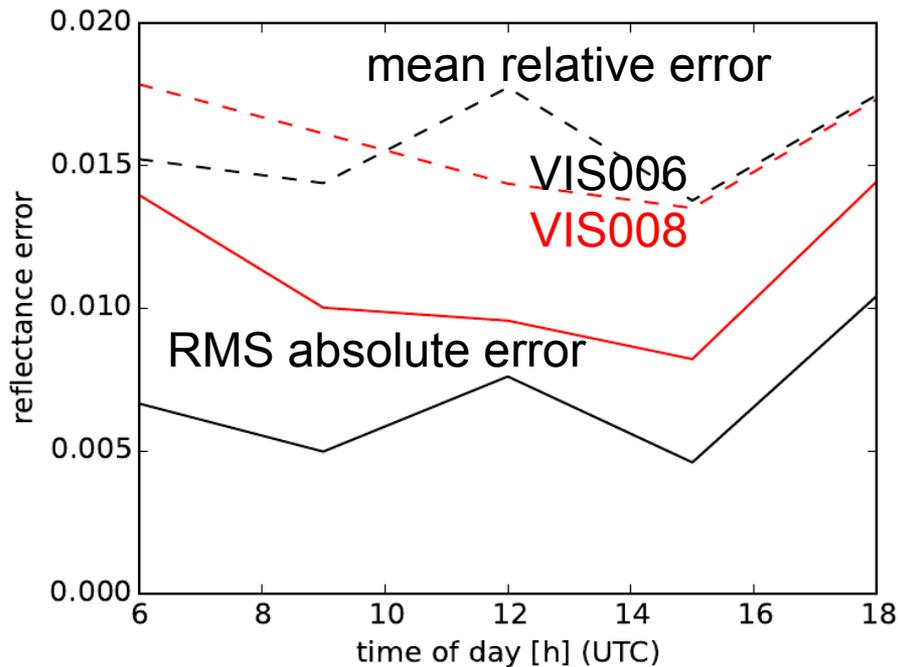
Compute **reflectance look-up table (LUT)** with discrete ordinate method (DISORT) for all parameter combinations

$\rightarrow$  effort for looking up reflectances: CPU-minutes

**Problem: Table is huge! O(10GB)**  $\rightarrow$  not suitable for online operator, slow interpolation  $\rightarrow$  **compress table to 20MB** using truncated Fourier series  $\rightarrow$  CPU-seconds

# Accuracy and computational effort

**Error of MFASIS (8 parameters/pixel) with respect to DISORT (full profiles available)**  
(model data: COSMO-DE fcsts for 10-28 June 2012)



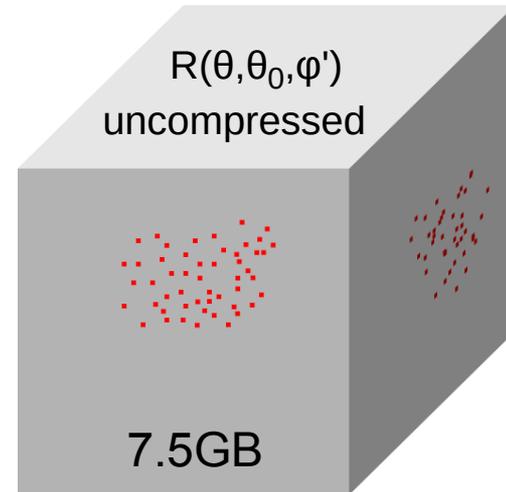
**Relative error < SEVIRI calibration error (~4%) for almost all pixels**

**Computational effort per column:**

DISORT (16 streams):  $2.3 \times 10^{-2}$  CPUsec

MFASIS (21MB table):  $2.5 \times 10^{-6}$  CPUsec

(on Xeon E5-2650, for 51 level COSMO data)



Impact of compression on performance?

Without compr.:  
LUT >> cache  
→ slow...

compression  
→ cache used efficiently



$R(\theta, \theta_0, \alpha)$ , compressed  
21MB

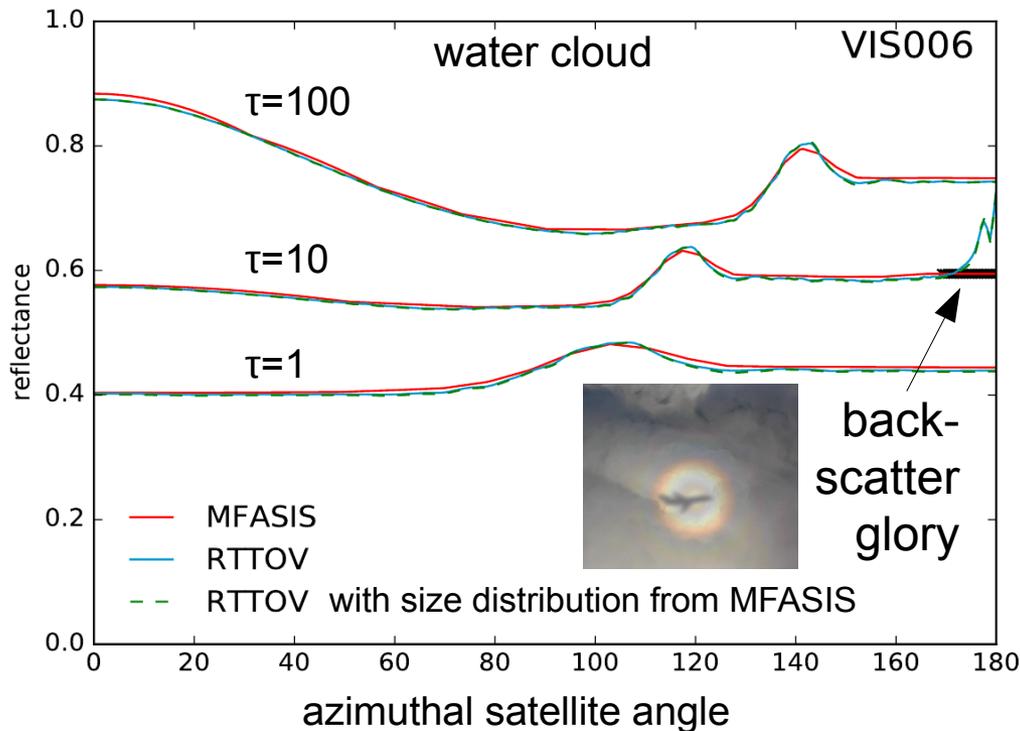
Scheck et al. 2016: *A fast radiative transfer method for the simulation of visible satellite imagery*, JQSRT, 175, pp. 54-67

## Comparison with RTTOV-DOM

(with J. Hocking, R. Saunders)

RTTOV-DOM: Implementation of DISORT in development at MetOffice / NWP-SAF

MFASIS & RTTOV-DOM were compared in the framework of DWDs NWP-SAF contribution



### Results:

- **Reflectances for clouds agree well!**
- Backscatter glory: reduced accuracy depends on unknown width of size dist.
- Clear sky contributions problems:
  - In MFASIS only a constant profile of water vapour is taken into account (affects the  $0.8\mu\text{m}$  channel)
  - Requires height-dependent reflectance correction (work in progress)
  - RTTOV-DOM: no multiple cloud - clear-sky scattering processes → negative reflectance bias

See [http://www.nwpsaf.eu/vs\\_reports/nwpsaf-mo-vs-054.pdf](http://www.nwpsaf.eu/vs_reports/nwpsaf-mo-vs-054.pdf)

## Improving accuracy and consistency

Having a fast and sufficiently accurate 1D RT solver is not enough...

**Errors sources in the operator**  
Approximations (e.g. missing 3D),  
Inconsistencies (e.g. subgrid clouds)  
missing information (→ spread)

**Errors in the NWP model state**  
e.g. cloud displacement (random),  
cloud cover bias (systematic)

**Errors in synthetic images**

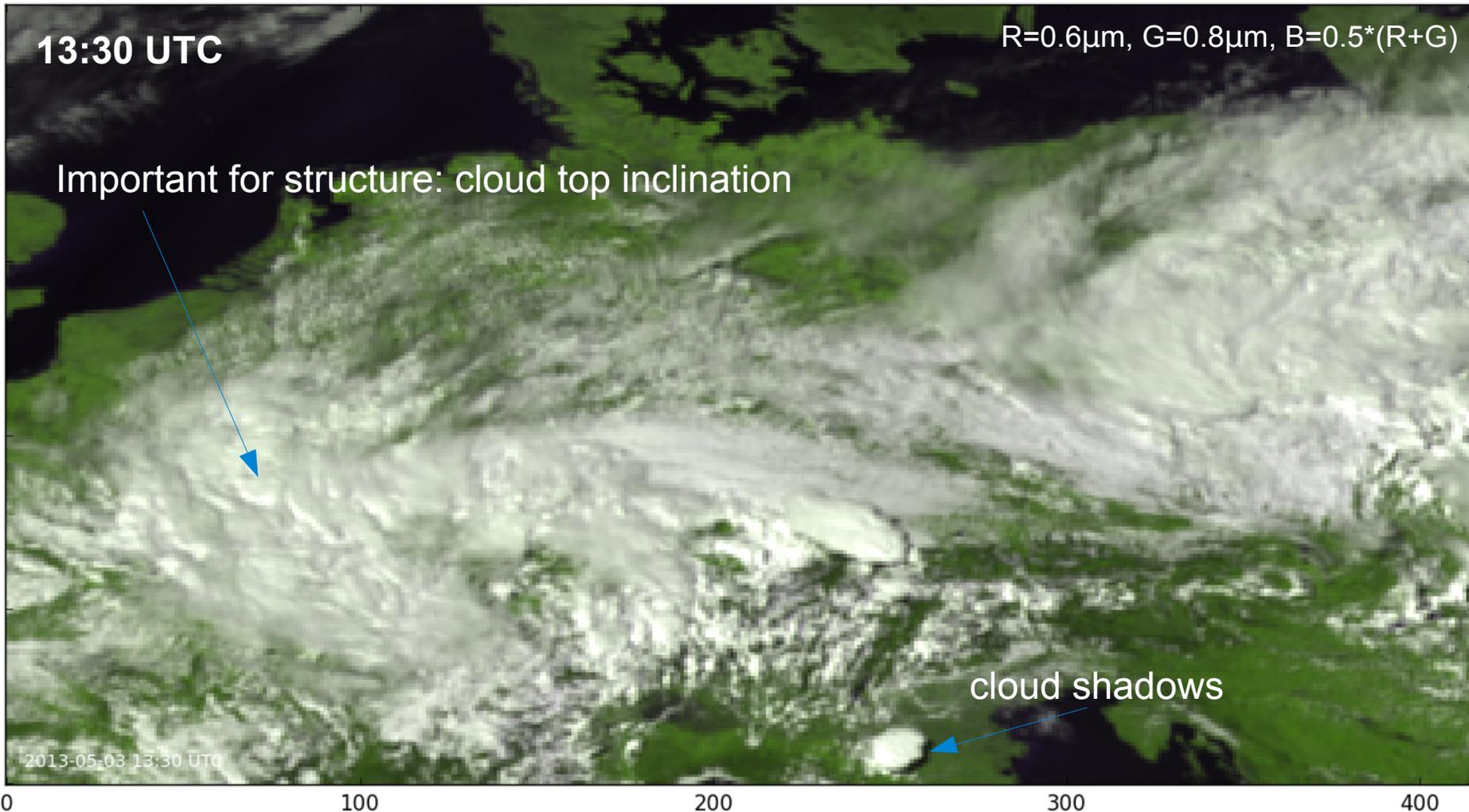
Operator error sources minimized / understood → model bias can be identified and removed → random errors can be reduced in DA

Most important features missing in operator version Kostka et al. 2014:

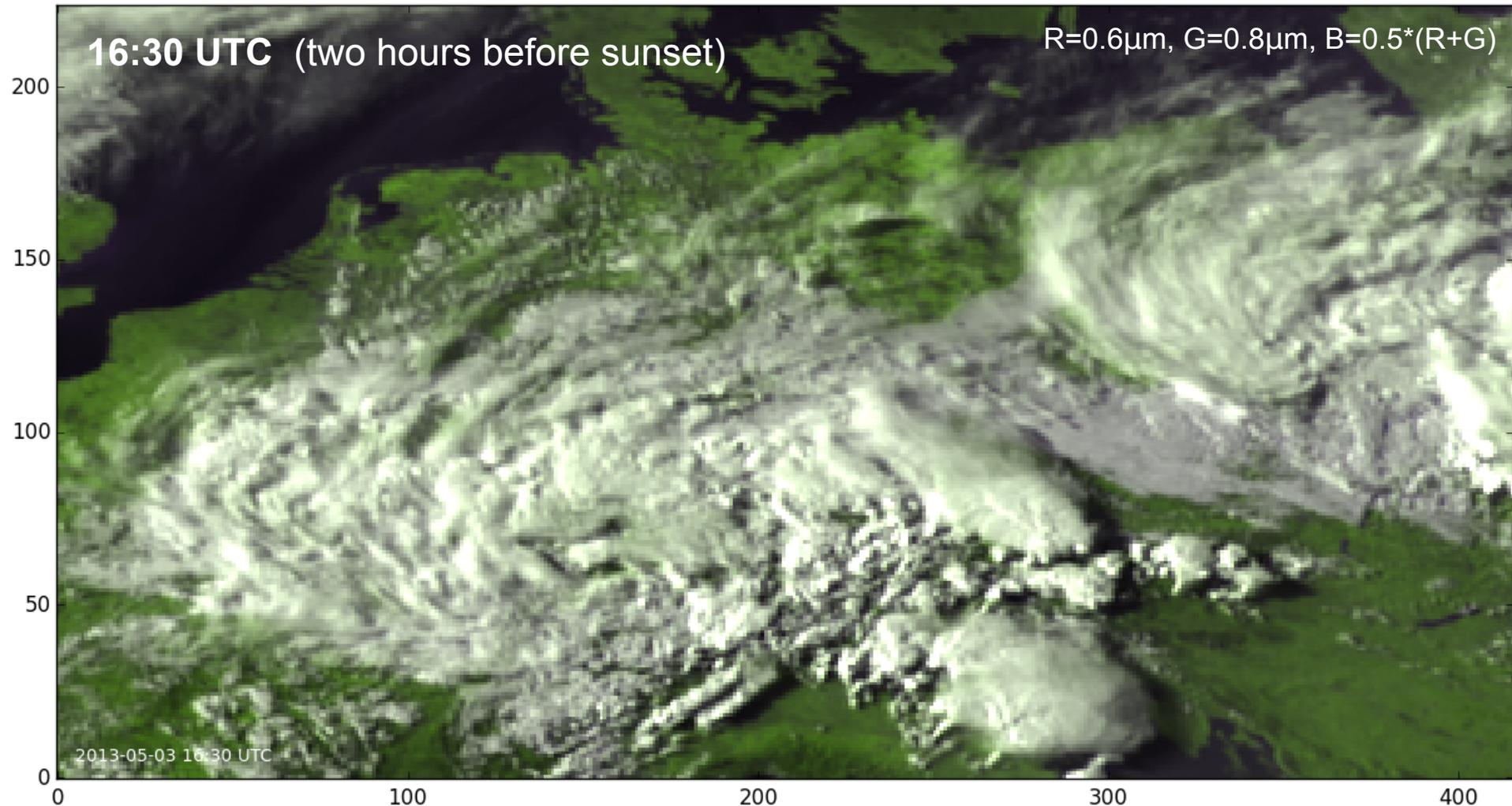
- **Cloud top inclination (3D RT effect)**
  - **Subgrid cloud overlap (consistency)**
- } How to take into account without degrading high performance?



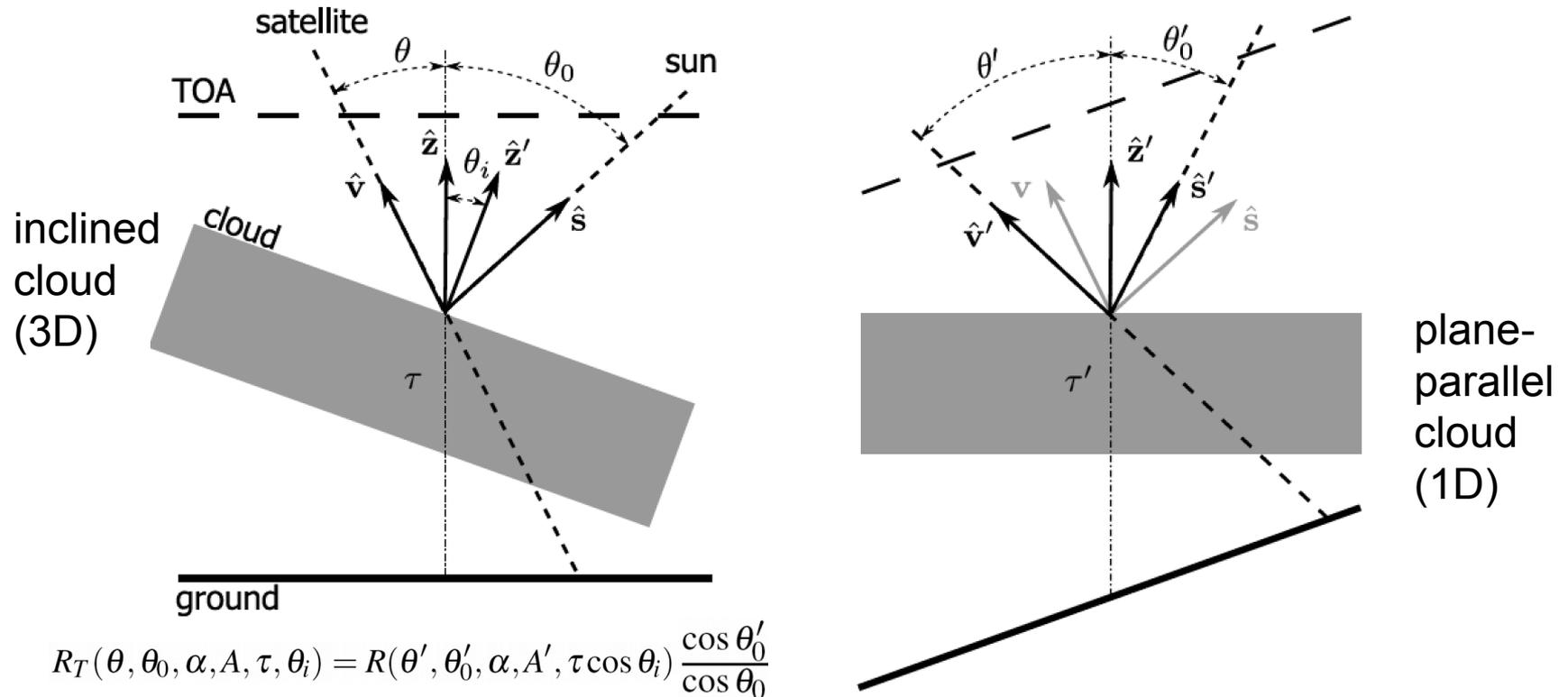
## 3D effects not accounted for in 1D radiative transfer



## 3D effects not accounted for in 1D radiative transfer

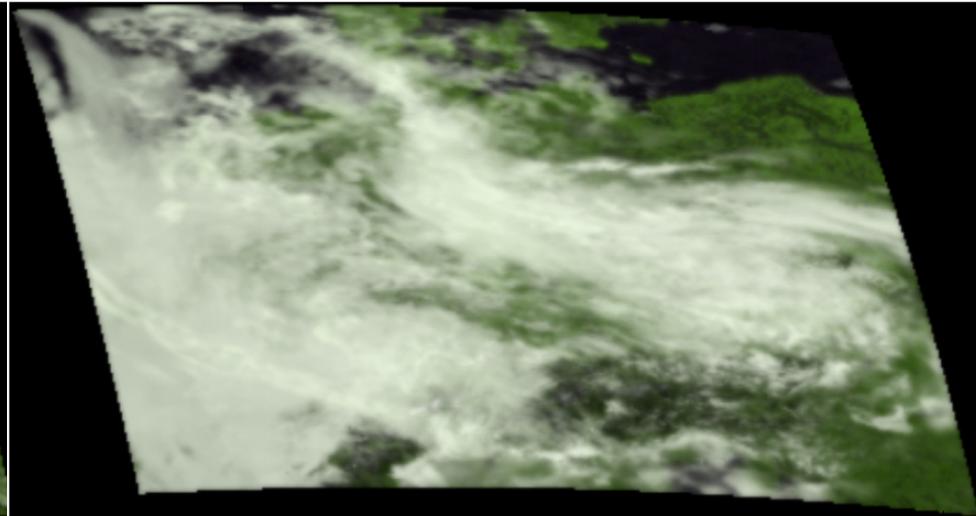
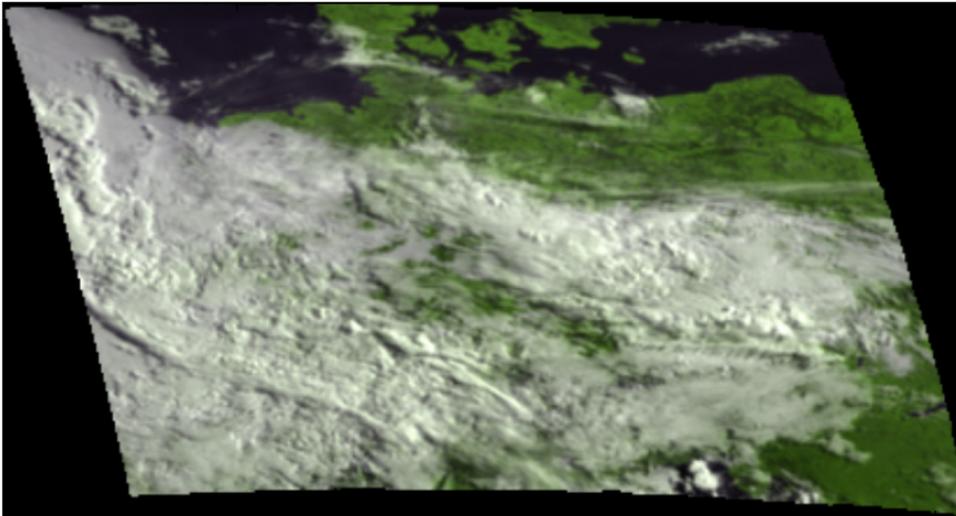


# Cloud top inclination correction



Rotated frame of reference with ground-parallel cloud → nearly a 1D problem (inclined ground is taken into account by using a modified surface albedo)  
→ Solve modified 1D problem, transform back to non-rotated frame.

## Cloud top inclination



SEVIRI 0.6 $\mu$ +0.8 $\mu$ , 3 June 2016, 6UTC

3h COSMO fcst **without 3D correction**

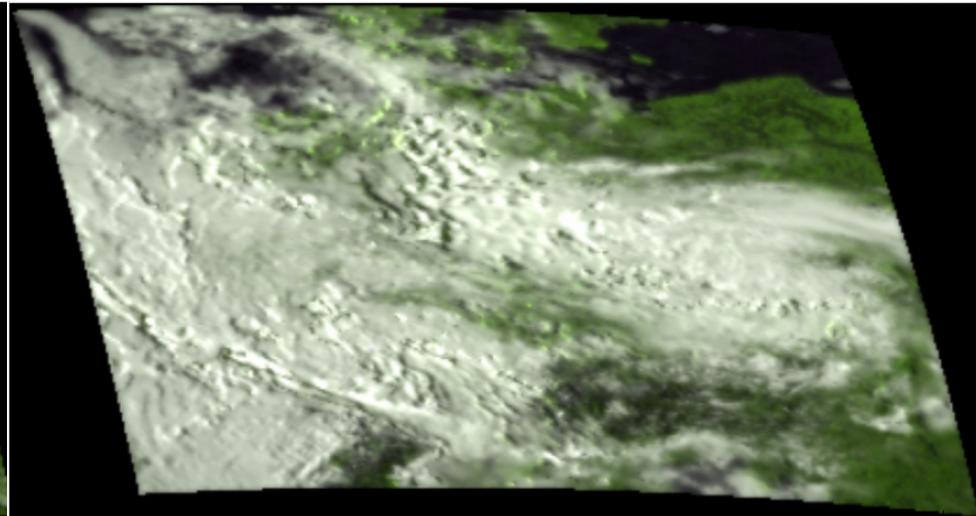
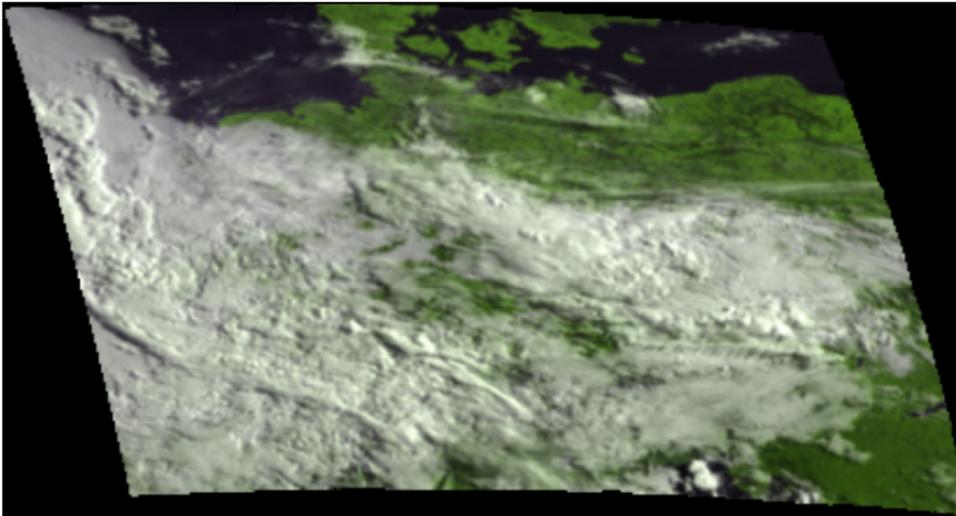
**Cloud top definition** : optical depth 1 surface  
(detect  $\tau=1$  in all columns, fit plane to column and 8 neighbour columns)

**Cloud top inclination correction** → **Increased information content**

Much more cloud structure is visible, in particular for larger SZAs

For instance, one can distinguish convective from stratiform clouds

## Cloud top inclination



SEVIRI 0.6 $\mu$ +0.8 $\mu$ , 3 June 2016, 6UTC

3h COSMO fcst **with 3D correction**

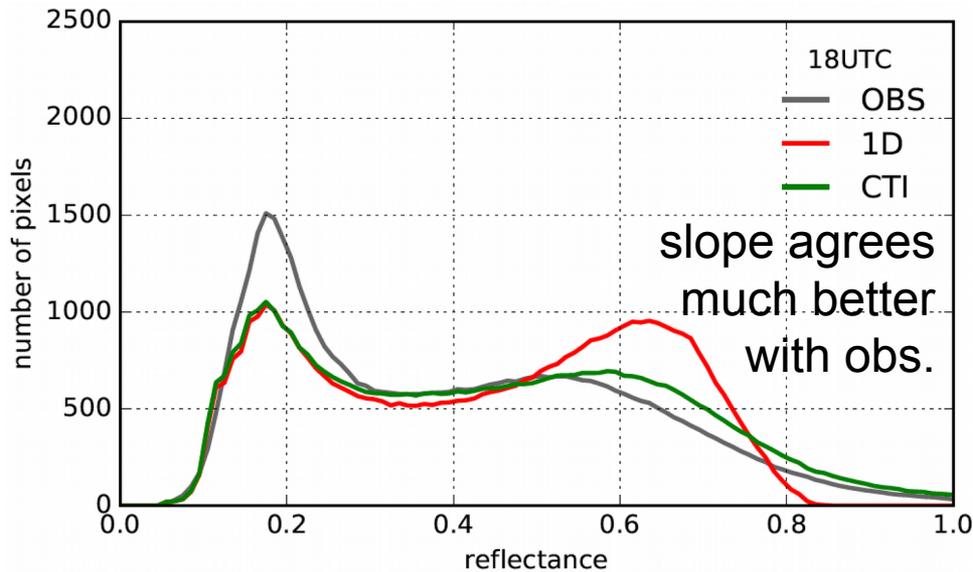
**Cloud top definition** : optical depth 1 surface  
(detect  $\tau=1$  in all columns, fit plane to column and 8 neighbour columns)

**Cloud top inclination correction** → **Increased information content**

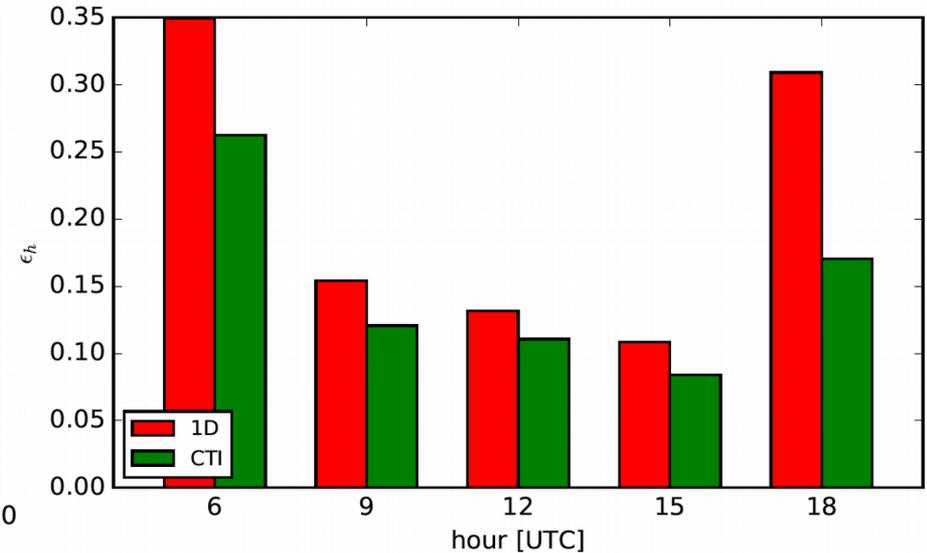
Much more cloud structure is visible, in particular for larger SZAs

For instance, one can distinguish convective from stratiform clouds

## Cloud top inclination correction



0.6 $\mu$ m reflectance histograms for 18UTC



area between obs.& model histogram

**Cloud top inclination correction** → **Systematic errors are reduced**

in particular for larger SZA, but some impact is always visible

**Computational effort: Small** (only  $\tau=1$  detection + one additional MFASIS call)

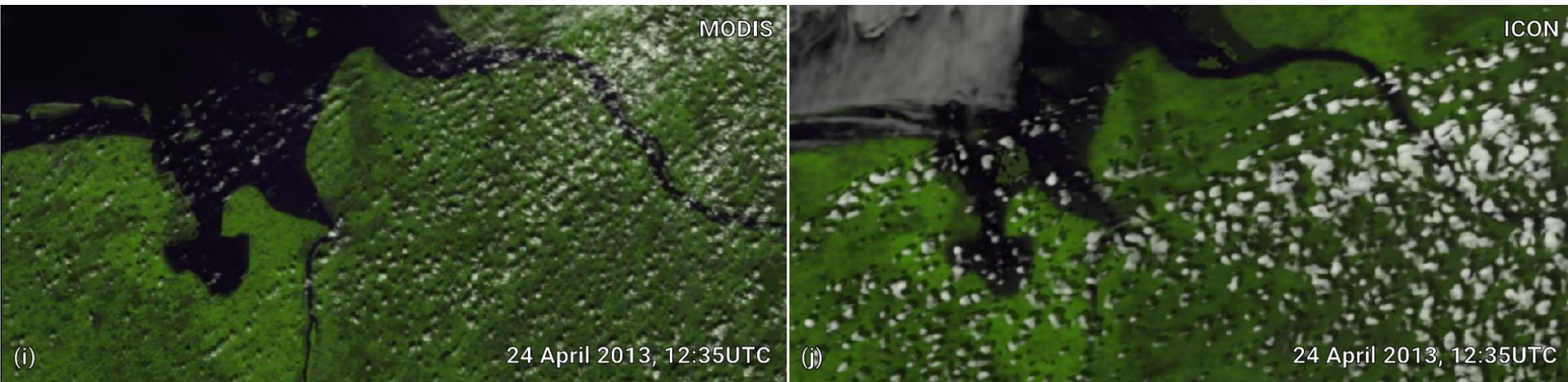
It should even be possible to include it in the real-time version (work in progress)

# MFASIS + 3D correction in real-time on GPUs

Master thesis by Theresa Diefenbach in the “Waves to Weather” project:  
MFASIS in Met3D (Marc Rautenhaus, TUM), runs interactively with ~10 frames/sec

System	Volume raycaster	
Scene 3	enabled	<input checked="" type="checkbox"/> True
Scene 4	configuration	
	rendering	
	wire frame	<input type="checkbox"/> False
	reload shaders	(click to ex...
	actor properties	
	▶ labels	
	render mode	MFASIS
	observed variable	clwc (fc)
	shading variable	clwc (fc)
	▼ mfasis visualization	
	effective radius	0.000010
	render IWC	<input type="checkbox"/> False
	load surface albedo map	(click to ex...
	map file	
	use MFASIS LUT	<input checked="" type="checkbox"/> True
	second pass	<input checked="" type="checkbox"/> True
	use Transferfunction	<input type="checkbox"/> False
	▼ isosurface raycaster	
	render Tau instead of LWC	<input type="checkbox"/> False
	▶ isovalues	
	▼ sampling step size	
	step size	0.003
	interaction step size	1.000
	bisection steps	4
	interaction bisection steps	4
	▶ shadow	
	▶ bounding box	
	▶ lighting	
	▶ normal curves	
	variables	

## A second 3D correction: Cloud shadows on the ground



Example: MODIS image + model equivalent for 150m resolution ICON run from HD(CP)<sup>2</sup> (see Heinze et al. (2017) “Large-eddy simulations over Germany using ICON”, QJRMS)

- Important for deep convection and broken cloud fields, in particular for  $0.8\mu\text{m}$
- Columns tilted towards sun  $\rightarrow$  shadow position. Brightness of shadows will often be dominated by diffuse radiation (problematic...)
- Preliminary implementation in operator version for the ICON model (parallel, offline or online), used for model evaluation (e.g. cloud size statistics)

## Subgrid cloud overlap

Common for NWP models: **Subgrid clouds** covering only a fraction of the grid cell are assumed to exist where relative humidity exceeds critical value.

Two or more partially cloudy cells in one column:  
**How do they overlap?** Affects heating, reflectance

**COSMO: Random-maximum overlap rules:**

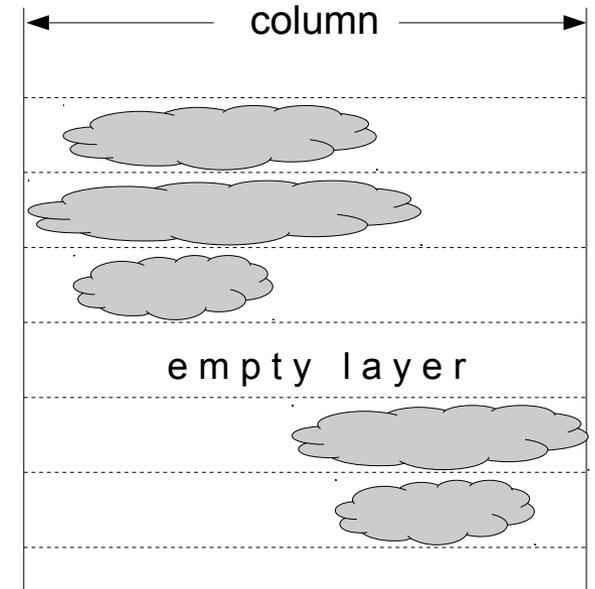
Clouds in adjacent layers overlap maximally, clouds separated by empty layers overlap randomly.

**Deterministic** schemes: Estimate mean reflectance of all allowed configurations

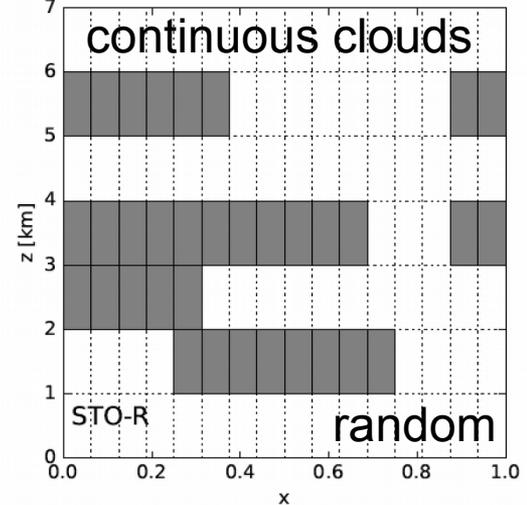
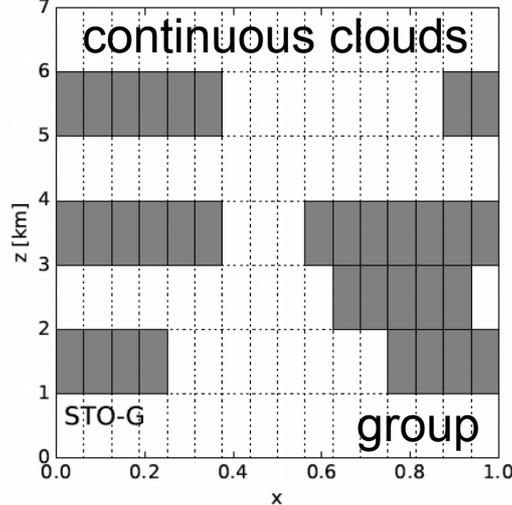
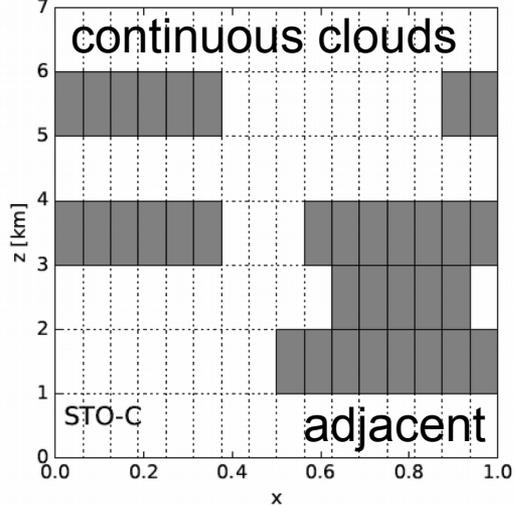
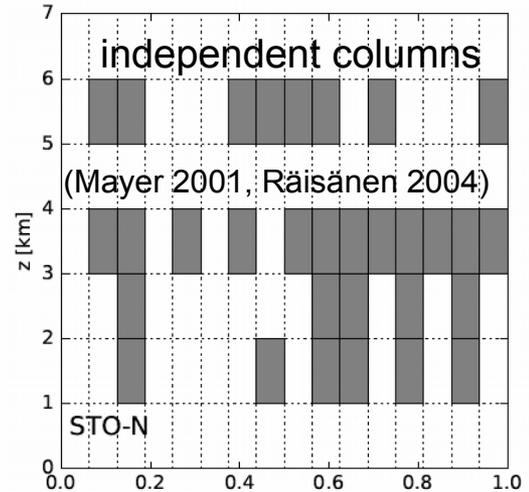
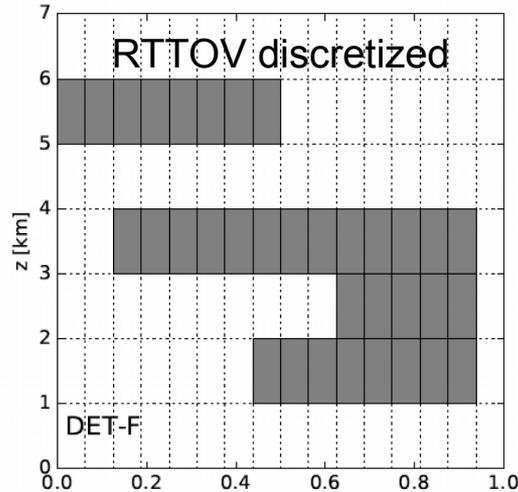
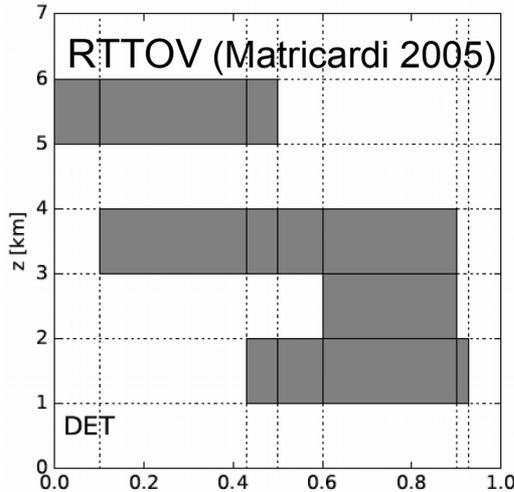
**Stochastic** schemes: Compute reflectance for one random realization  
(spread quantifies uncertainty in cloud distribution)

**Several schemes were compared to address these questions:**

- How well do different deterministic and stochastic implementations agree?
- Is the spread large enough to be relevant for DA?
- Should the slant viewing path of the satellite be taken into account?

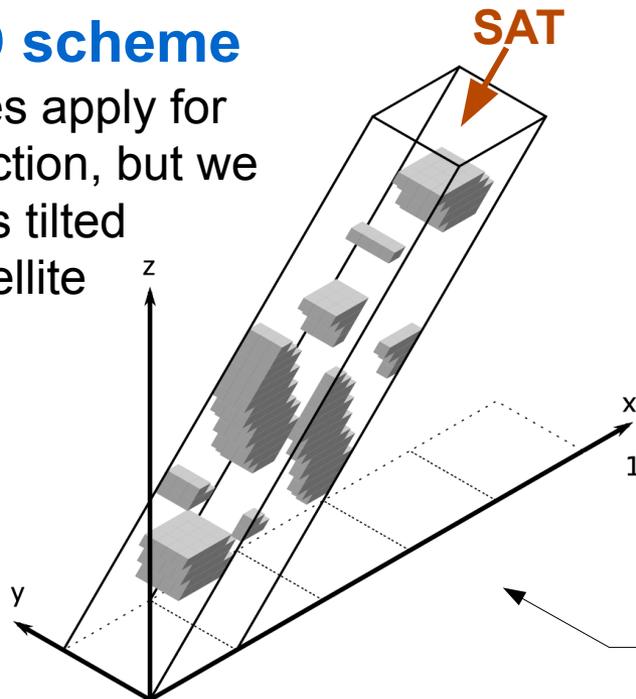
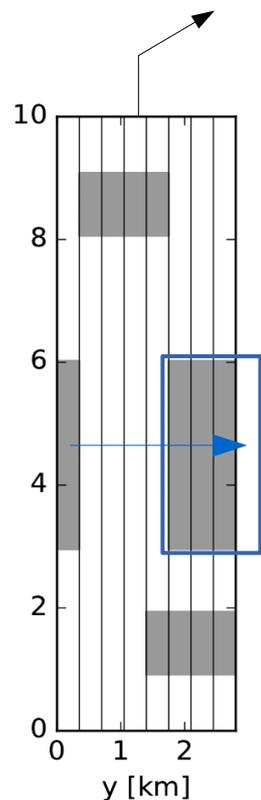


**Common strategy:** Subdivide column, fill subgrid cells according to overlap rules (different cloud size dist. possible), perform RT for each subcolumn, average results



## A new 3D scheme

Overlap rules apply for vertical direction, but we use columns tilted towards satellite

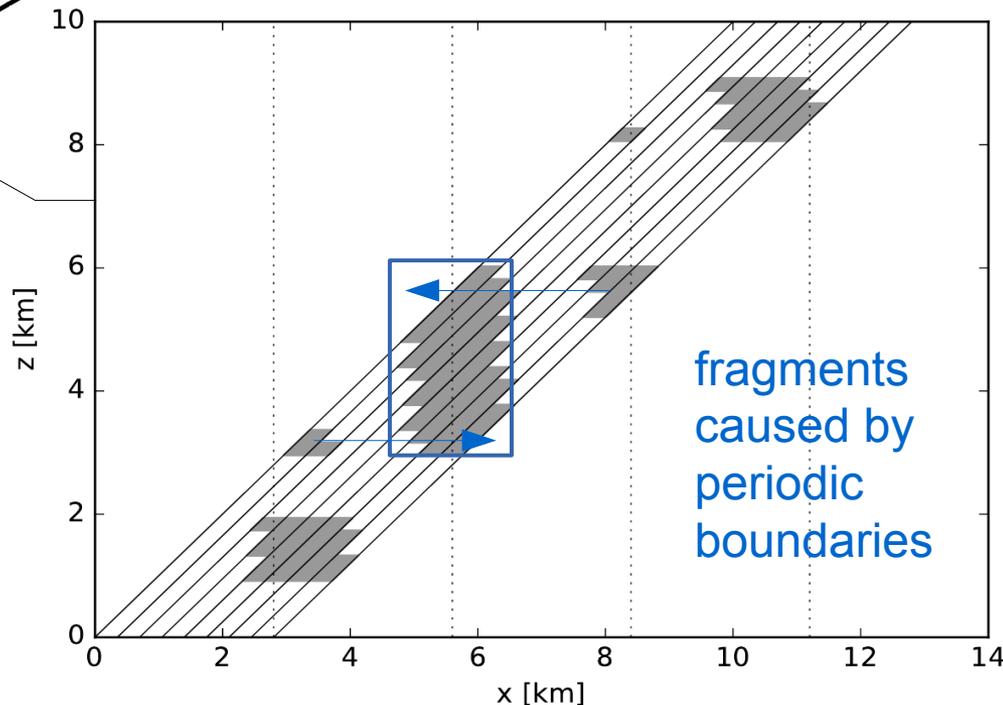


→ **Increased total cloud cover**  
(also cloud sides contribute)

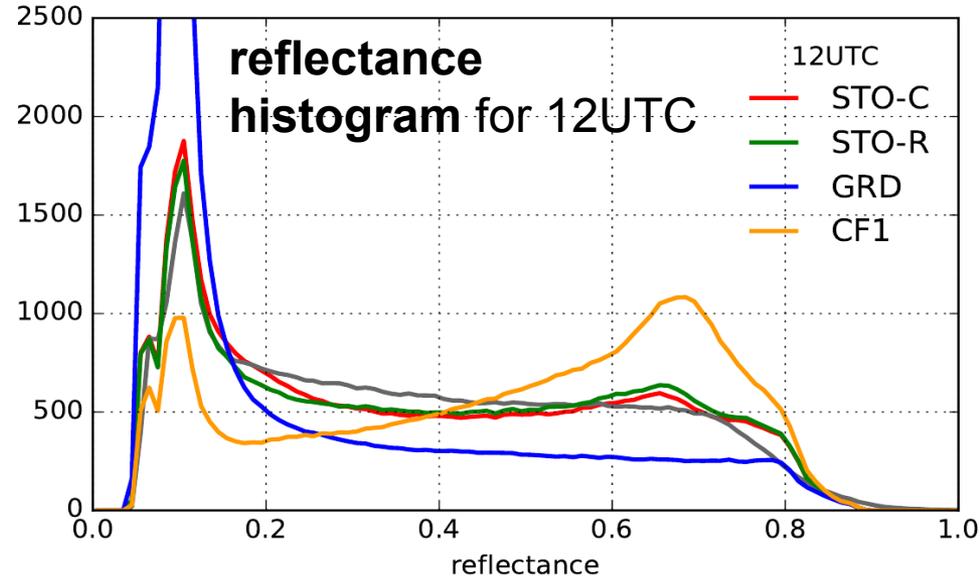
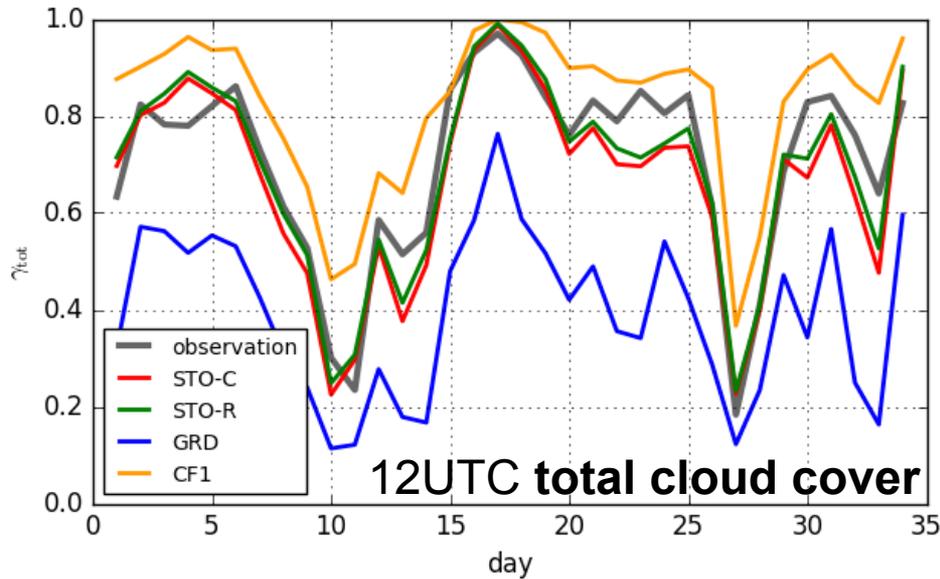
Not more expensive than 2D schemes.

**Approach:** Use bundle of  $N \times N$  sub-columns (3D), compensate for slant viewing path by shifting clouds into x-direction in each layer

**Example:** 3 clouds with constant cloud fraction 0.25 spanning several layers  
→ vertical clouds, consistent with model



# Results for operational COSMO forecasts in June 2016



**It is essential to take cloud overlap into account,** setting all clouds fractions to 1 or using only grid scale clouds causes large errors.

Differences related to different assumptions or implementations are much smaller.

**Good agreement with observations (no tuning!)**

SEVIRI observation

grid scale clouds only

Subgrid cloud fraction 1

random overlap

random-maximum overlap

(2D stochastic continuous clouds)

## Overlap schemes: local impact

### Random vs. rand.-max. overlap

Local impact can be significant, ensemble mean random - randmax can be  $> 0.1$ , i.e. several 10%, but only  $\sim 10\%$  of the pixels are sensitive to the type of the overlap assumptions

### Random-maximum Implementations

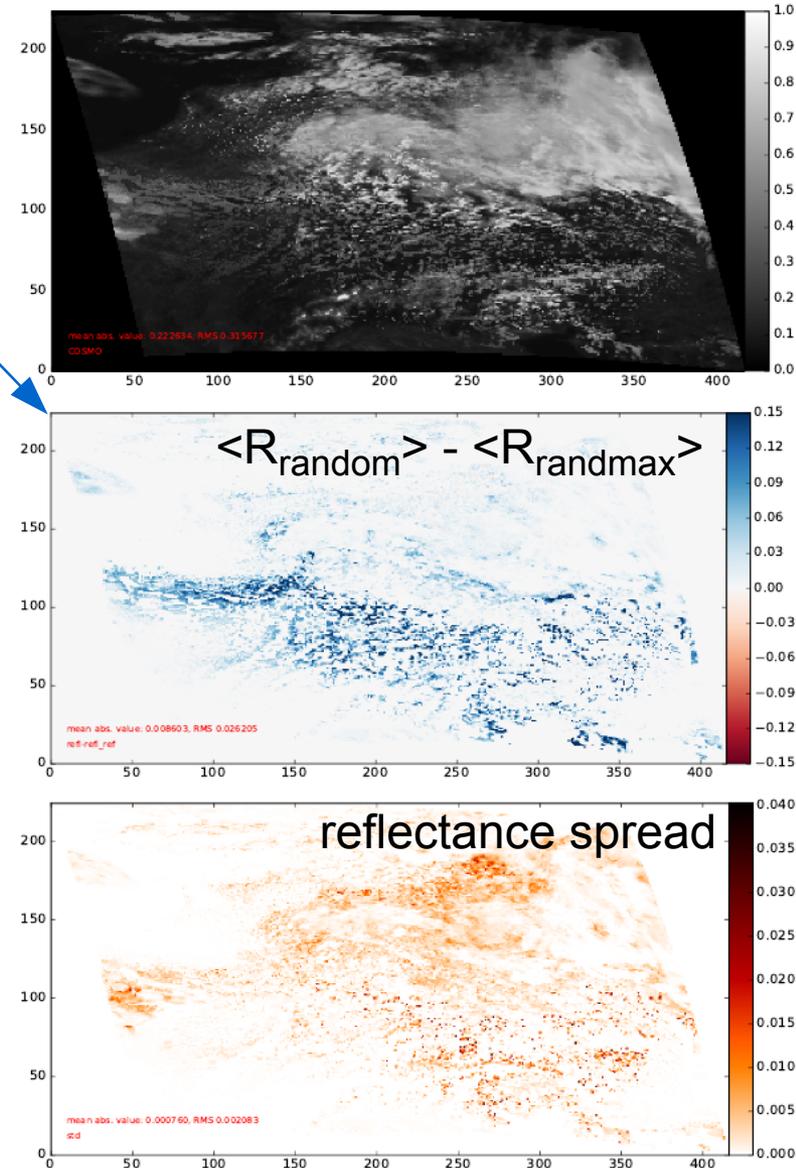
Mean reflectances of 2D stochastic schemes are very similar, also to deterministic schemes.  $\sim 10$  subcolumns are sufficient.

### Consistency

Taking slant viewing angle into account (3D) has same impact as switching rand./max.  $\rightarrow$  random (at latitude  $\sim 45^\circ$ , stronger effect for higher lats.)

### Spread

Small, spread  $> 0.01$  only in  $\sim 15\%$  of pixels  $\rightarrow$  should not have significant impact on DA



## First assimilation results

Assimilation of conventional and/or SEVIRI obs. in COSMO/KENDA

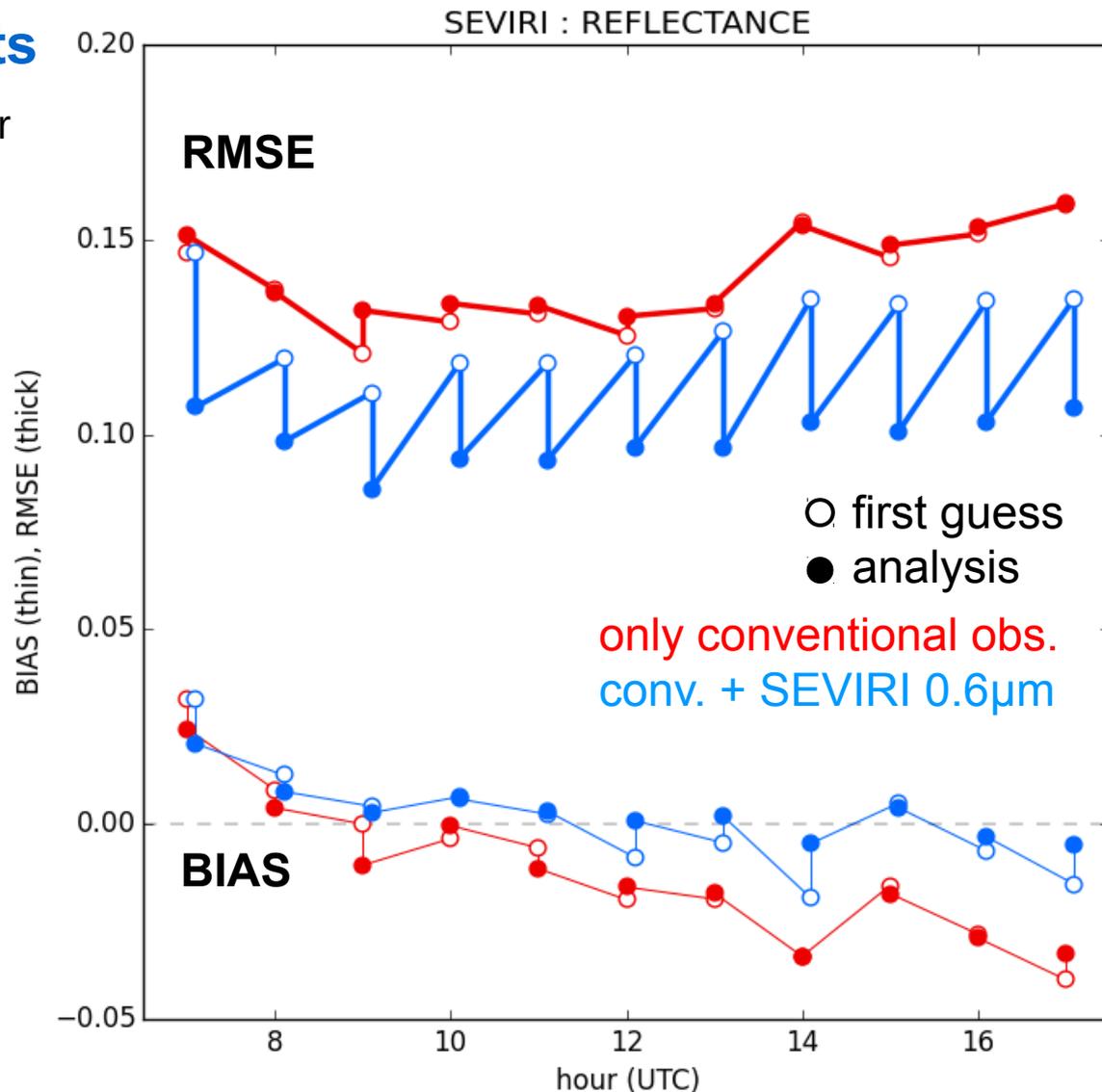
### Setup:

40 member LETKF  
1h assimilation interval  
0.6 $\mu$ m observations  
Observation error 0.2  
Superobbing (radius 3 pixels)  
Horiz. localization 100km  
No vertical localization

Assimilation of SEVIRI observations:

**lower reflectance**  
**RMSE and bias**

Independent GPS humidity observations: reduced error



## Summary

- ➔ Visible & near-infrared channels could provide useful information for convective scale DA
- ➔ We have developed MFASIS, a 1D RT method that is sufficiently fast for operational DA
- ➔ The most important 3D RT effect is related to the inclination of cloud tops and can be taken into account approximately in a efficient way
  - increased information content,
  - reduced systematic error
- ➔ Overlap of subgrid clouds is important. Most consistent scheme takes slant satellite viewing path into account.
- ➔ First assimilation experiments with DWD KENDA (LETKF) are promising, more experiments with new operator version will be performed soon...

