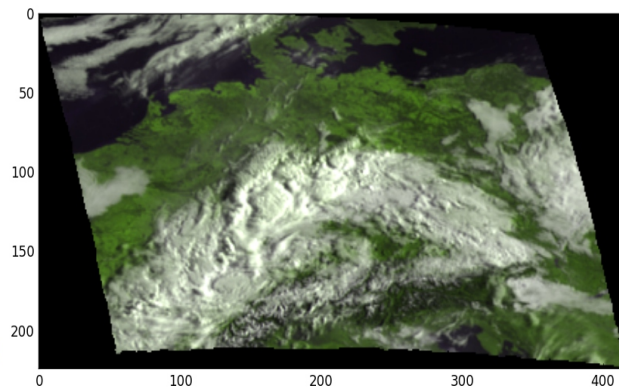
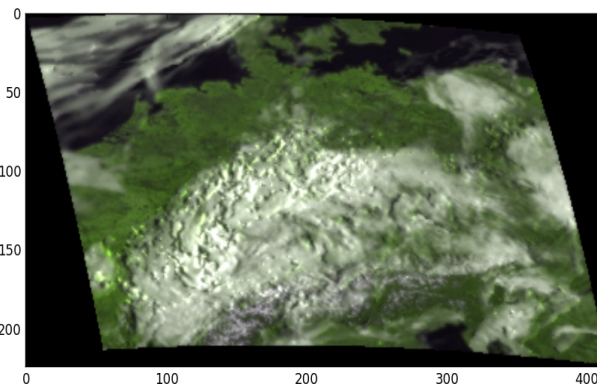
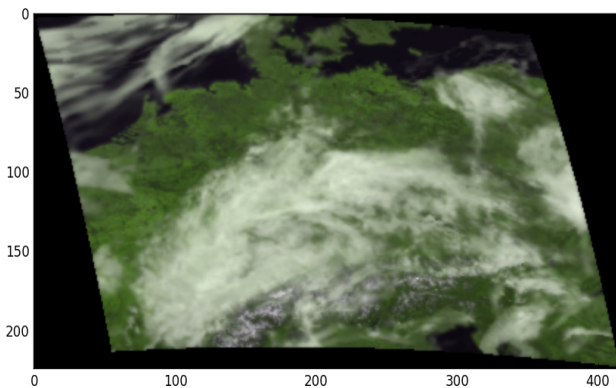


Using visible and near-infrared satellite observations for convective-scale data assimilation

Leonhard Scheck^{1,2}, Bernhard Mayer², Martin Weissmann^{1,2}

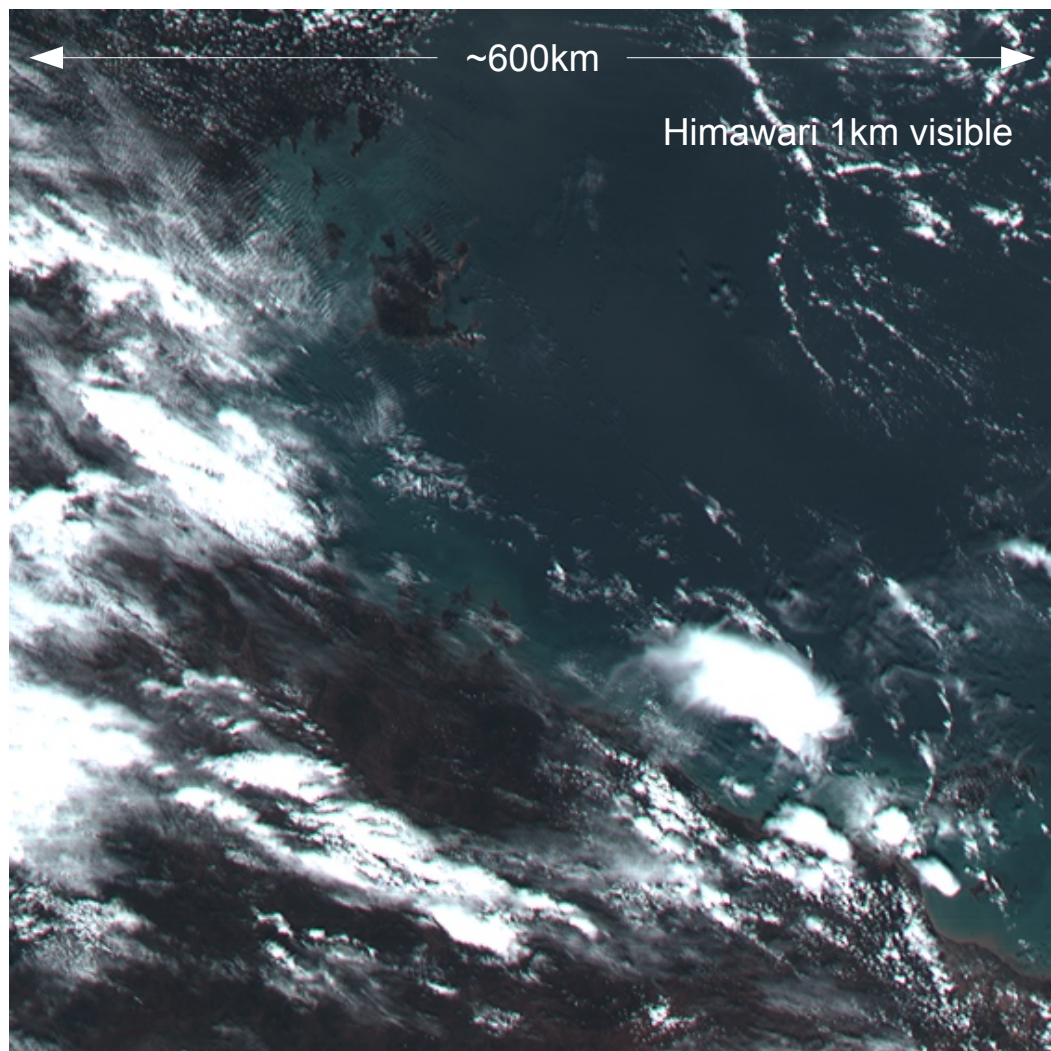
1) Hans-Ertel-Center for Weather Research, Data Assimilation Branch

2) Ludwig-Maximilians-Universität, Munich



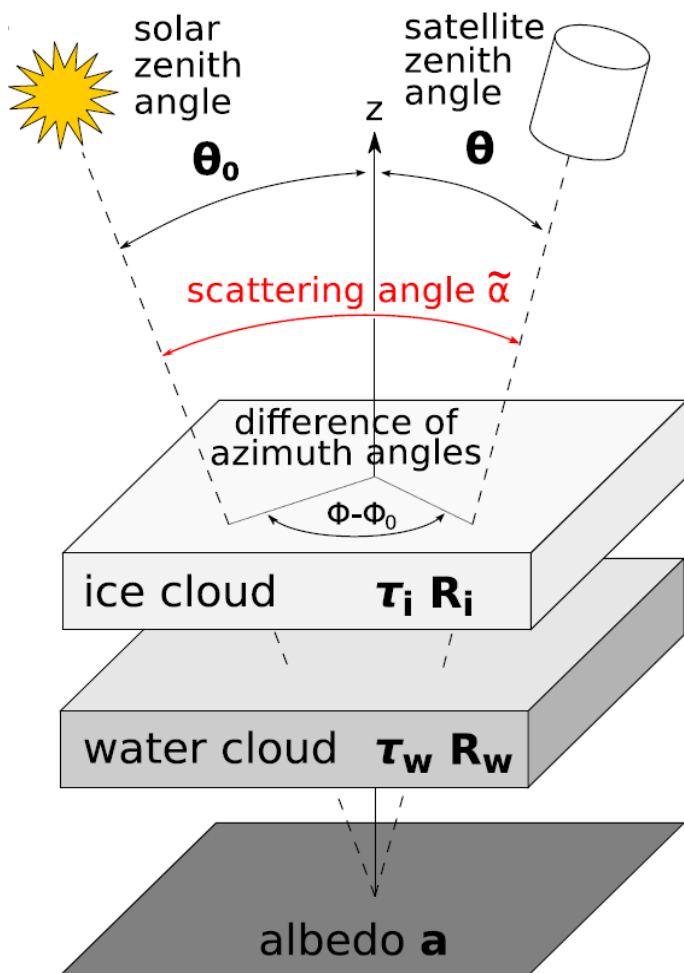
Visible / near-infrared satellite observations for DA

- relevant for **convective scale DA**:
high spatial and temporal resolution.
Himawari-8/9, GOES-R, MTG:
0.6 μ m resolution: 500m (IR: 2km)
6-8 of 16 channels $\lambda < 4\mu$ 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

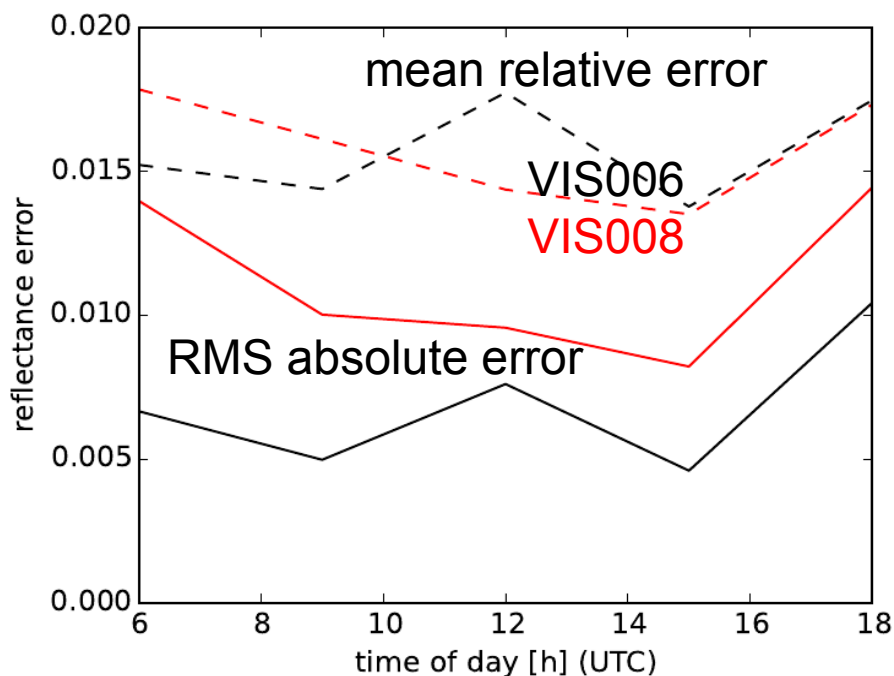
Compute **reflectance look-up table (LUT)** with discrete ordinate method (DISORT) for all parameter combinations

\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

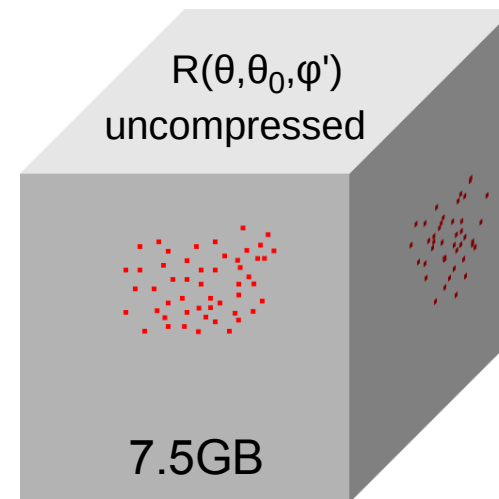
Accuracy and computational effort

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): 2.3×10^{-2} CPUsec
MFASIS (21MB table): 2.5×10^{-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

CPU cache
20MB



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

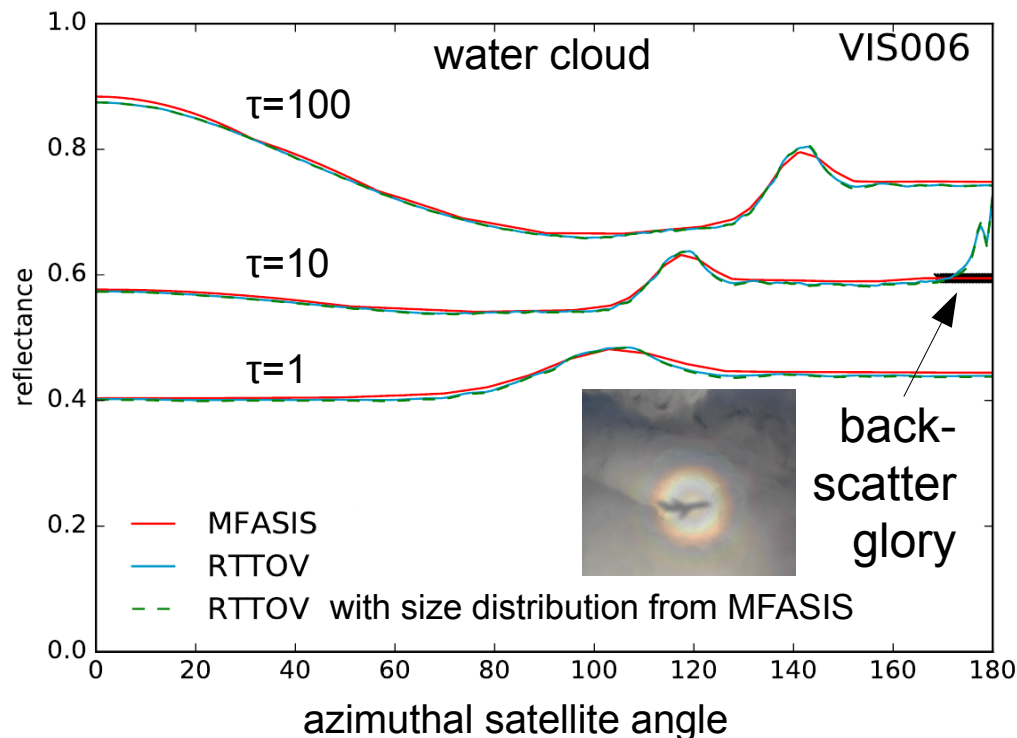
Scheck et al. 2016: *A fast radiative transfer method for the simulation of visible satellite imagery*, JQSRT, 175, pp. 54-67

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



See http://www.nwpsaf.eu/vs_reports/nwpsaf-mo-vs-054.pdf

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

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)

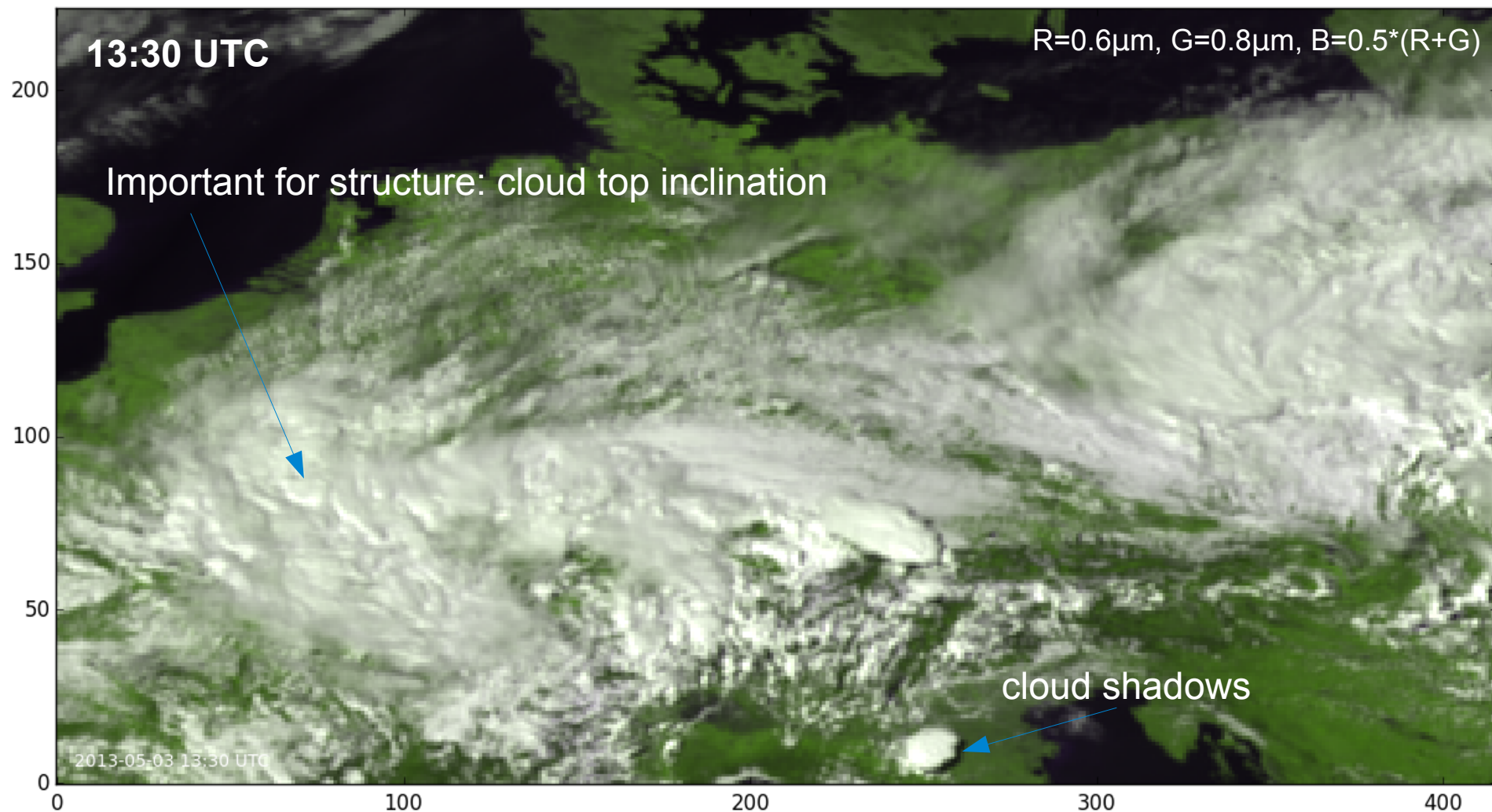
Errors in synthetic images

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

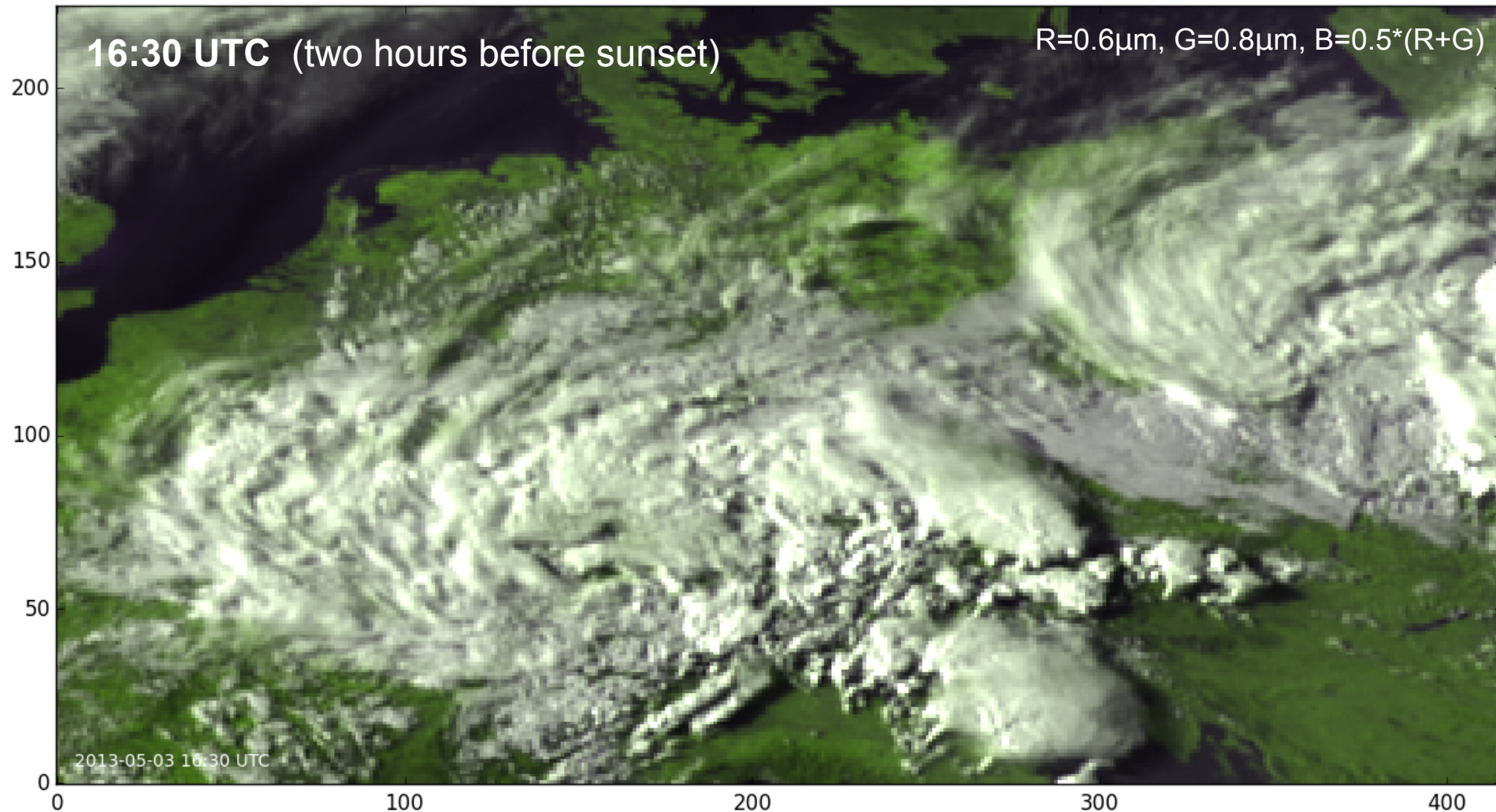
Most important features missing in operator version Kostka et al. 2014:

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

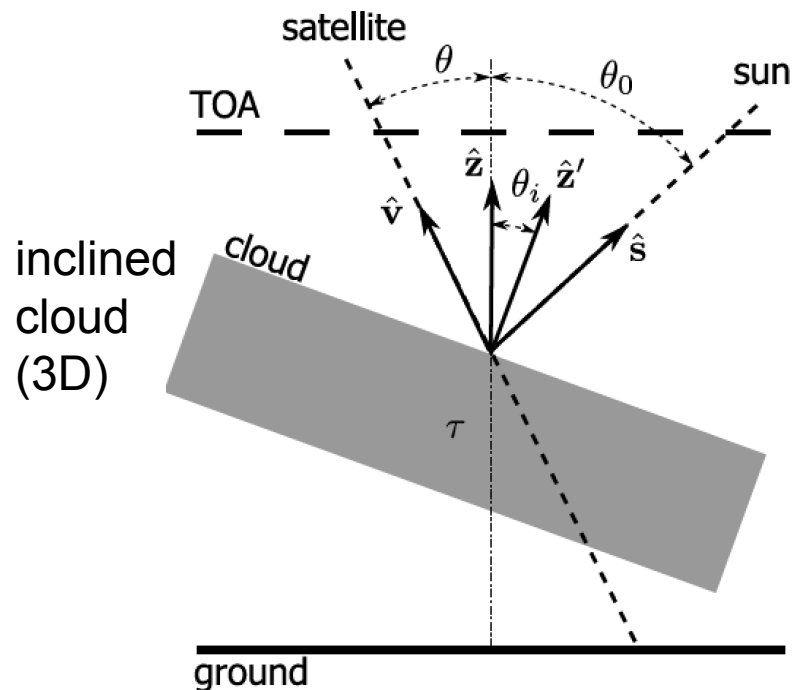
3D effects not accounted for in 1D radiative transfer



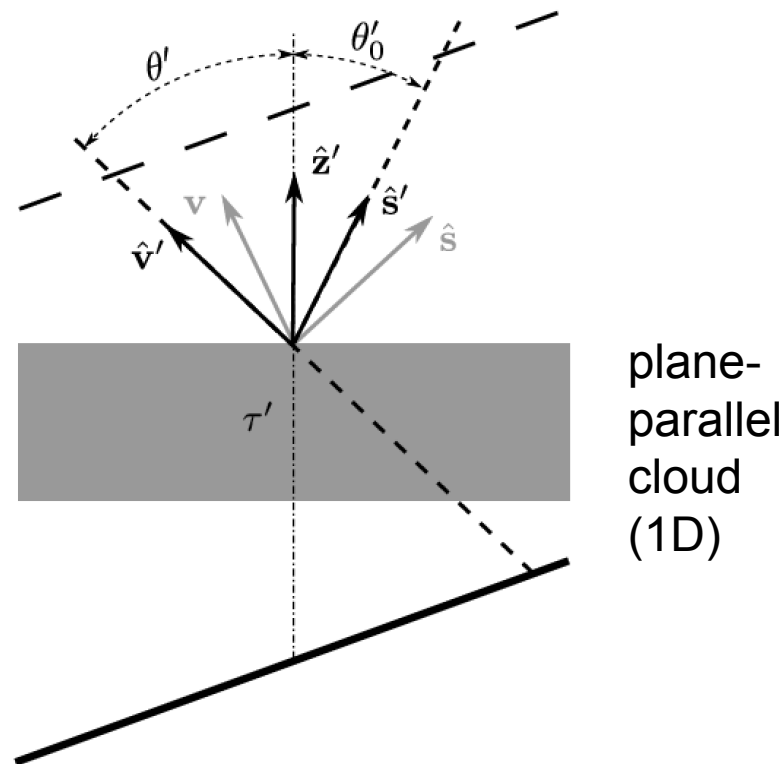
3D effects not accounted for in 1D radiative transfer



Cloud top inclination correction

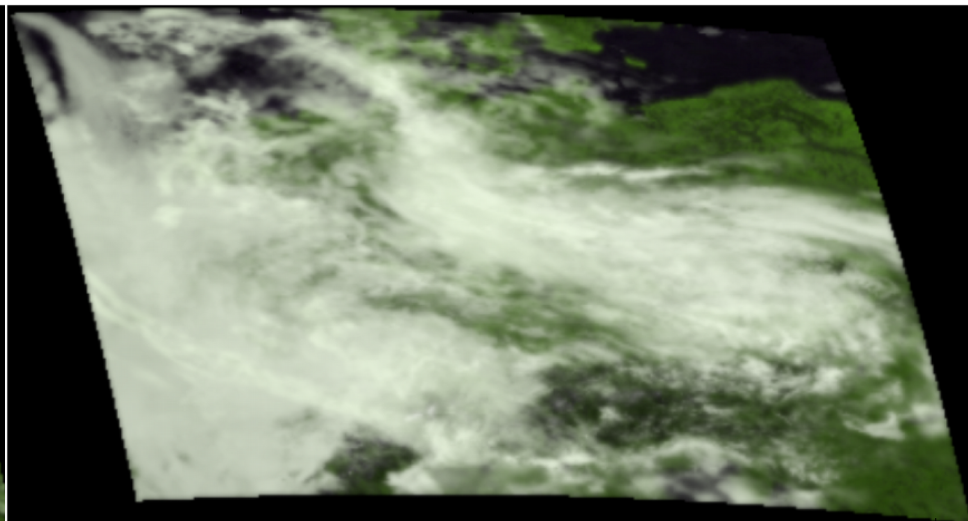
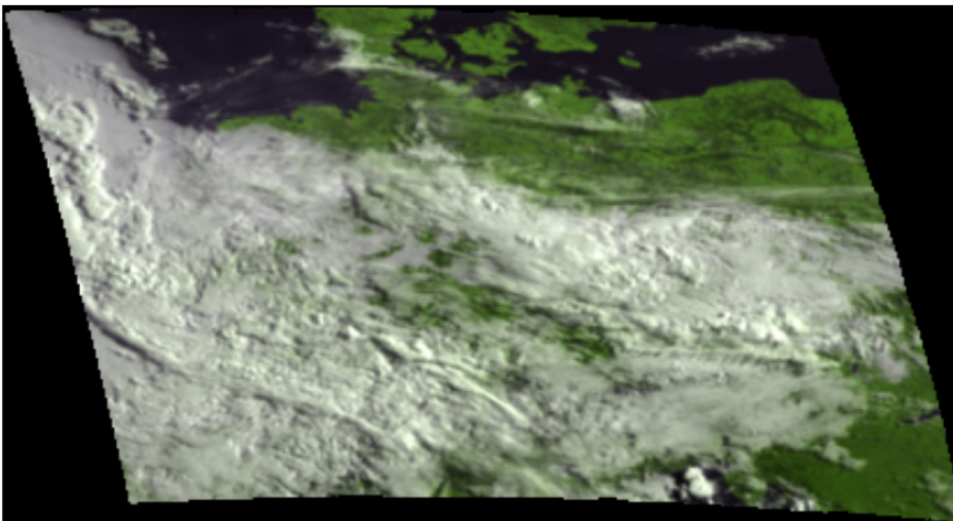


$$R_T(\theta, \theta_0, \alpha, A, \tau, \theta_i) = R(\theta', \theta'_0, \alpha, A', \tau \cos \theta_i) \frac{\cos \theta'_0}{\cos \theta_0}$$



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 μ +0.8 μ , 3 June 2016, 6UTC

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

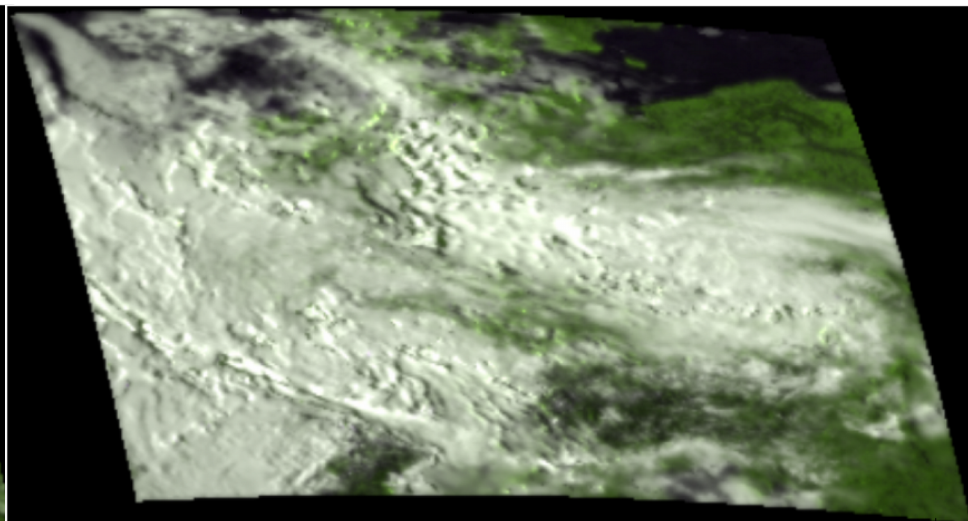
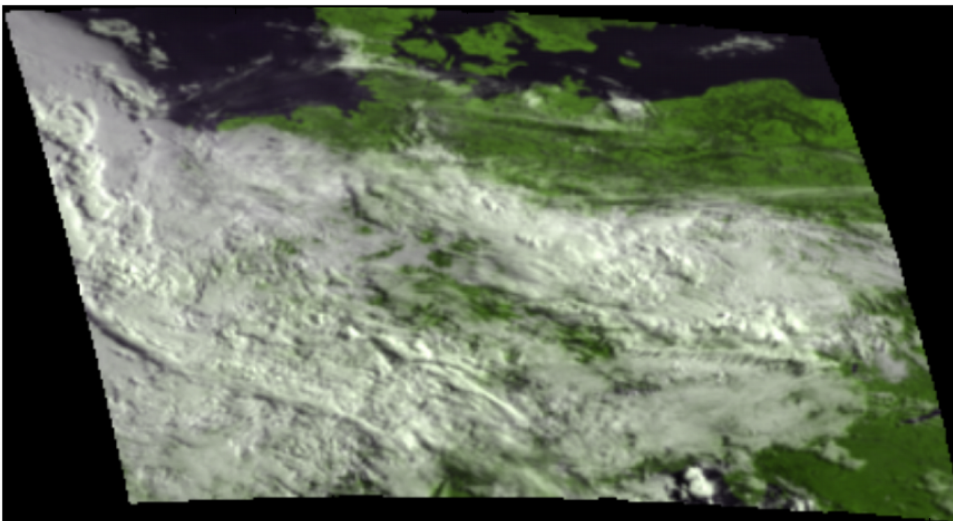
Cloud top definition : optical depth 1 surface
(detect $\tau=1$ in all columns, fit plane to column and 8 neighbour columns)

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 μ +0.8 μ , 3 June 2016, 6UTC

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

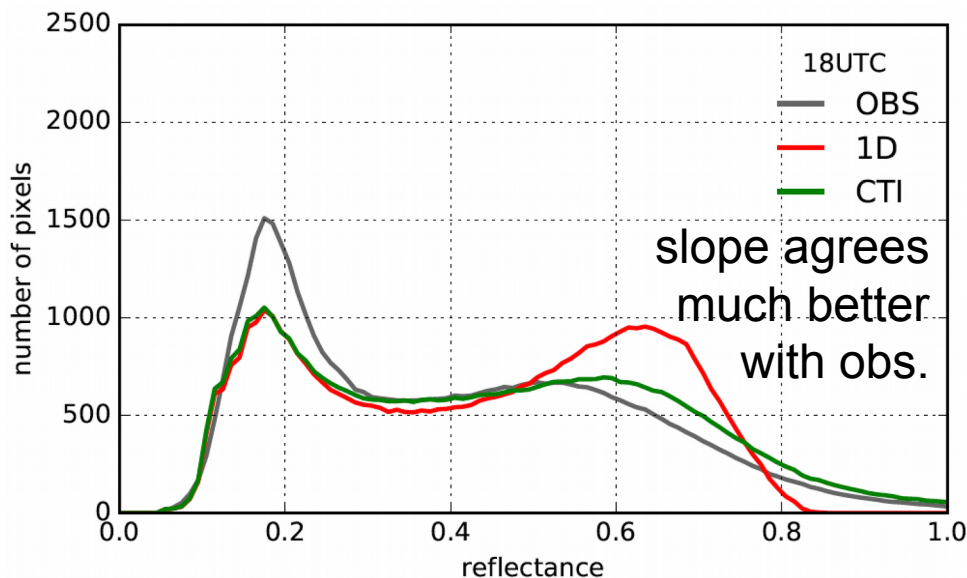
Cloud top definition : optical depth 1 surface
(detect $\tau=1$ in all columns, fit plane to column and 8 neighbour columns)

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

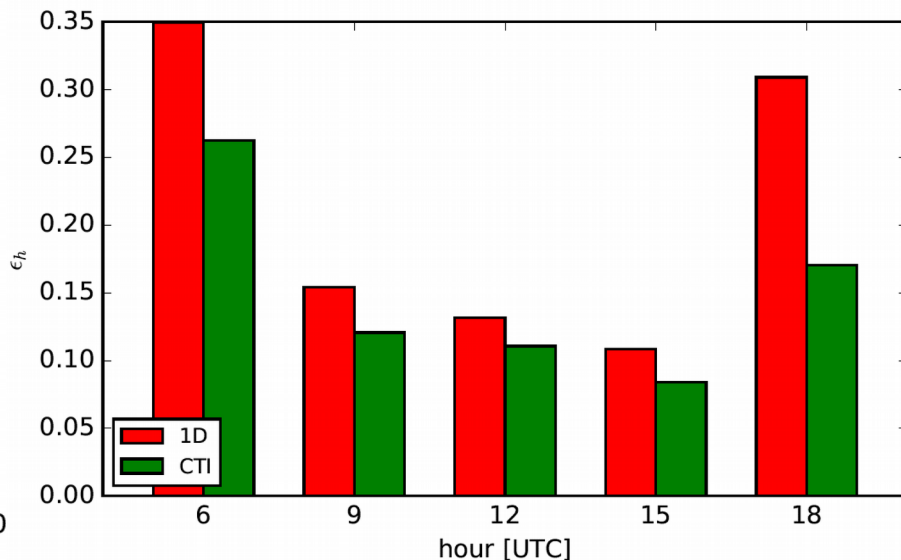
Much more cloud structure is visible, in particular for larger SZAs

For instance, one can distinguish convective from stratiform clouds

Cloud top inclination correction



0.6 μ m reflectance histograms for 18UTC



area between obs.& model histogram

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

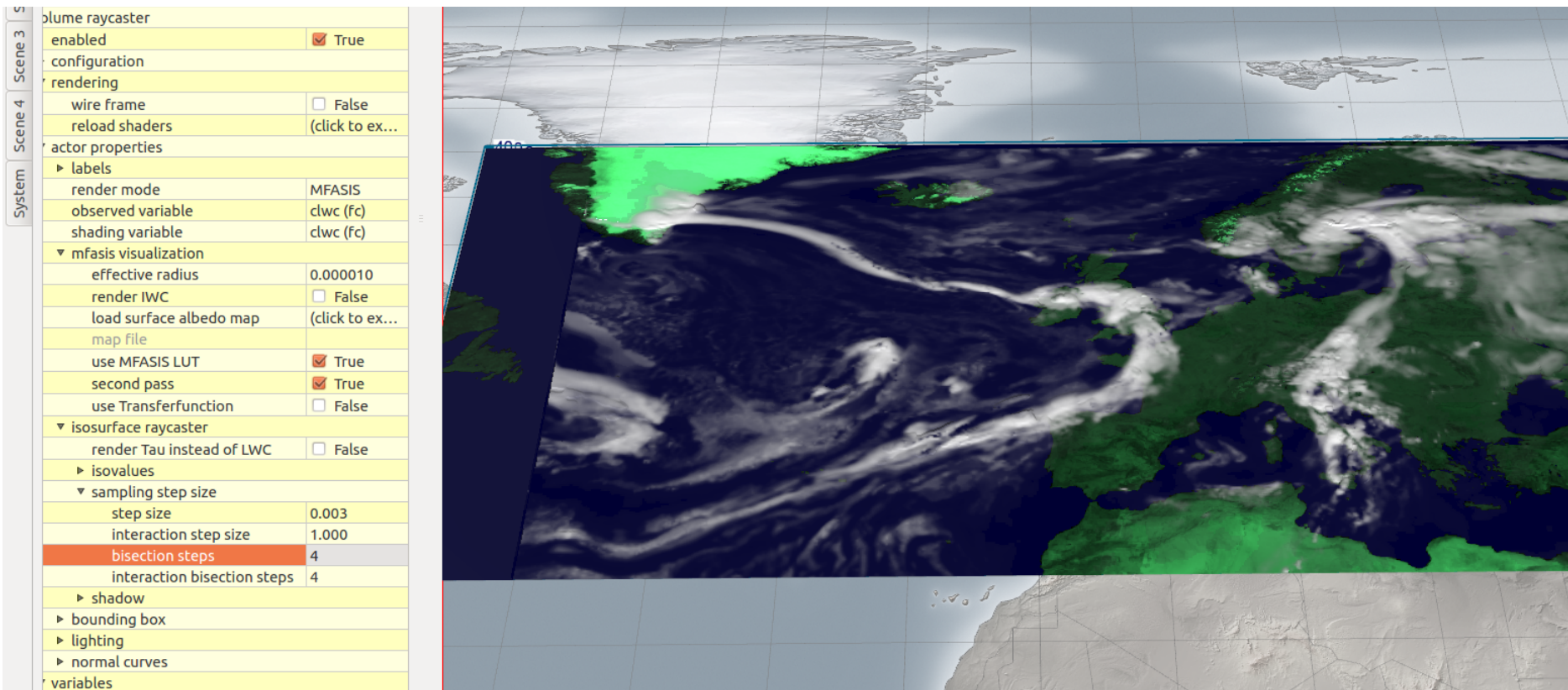
in particular for larger SZA, but some impact is always visible

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

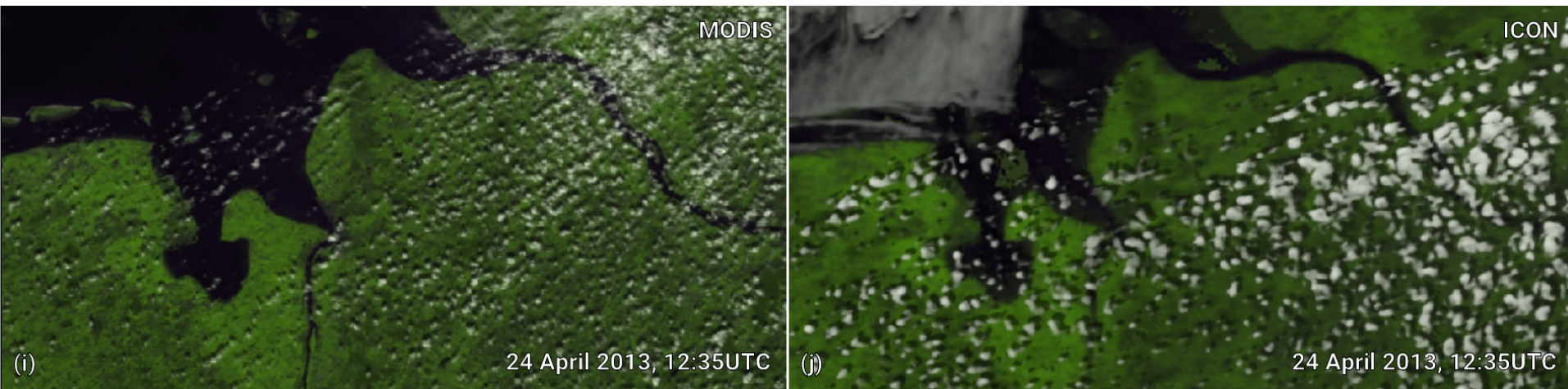
It should even be possible to include it in the real-time version (work in progress)

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



A second 3D correction: Cloud shadows on the ground



Example: MODIS image + model equivalent for 150m resolution ICON run from HD(CP)² (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 → 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:**

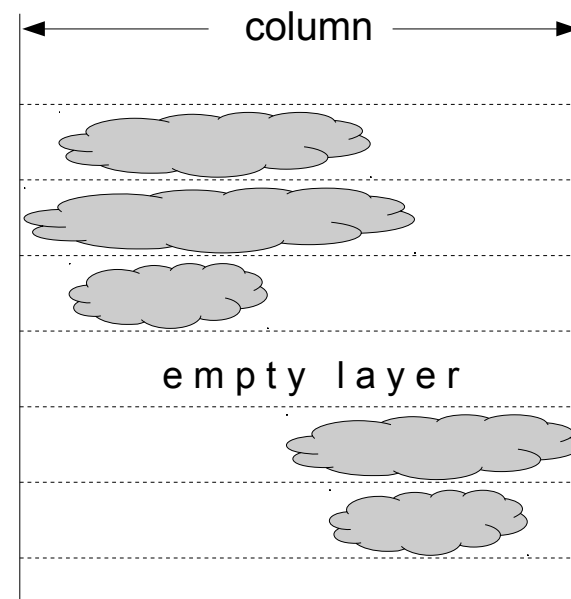
Clouds in adjacent layers overlap maximally, clouds separated by empty layers overlap randomly.

Deterministic schemes: Estimate mean reflectance of all allowed configurations

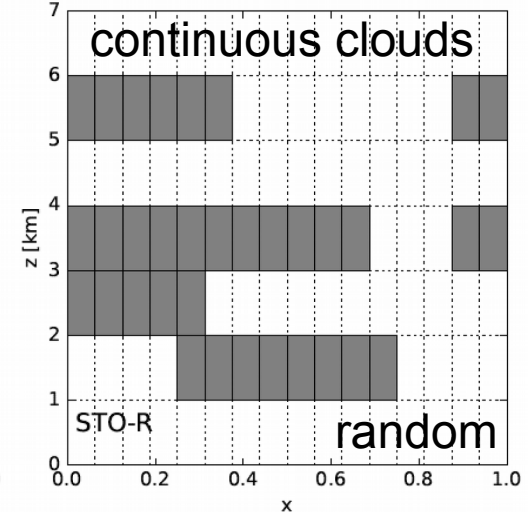
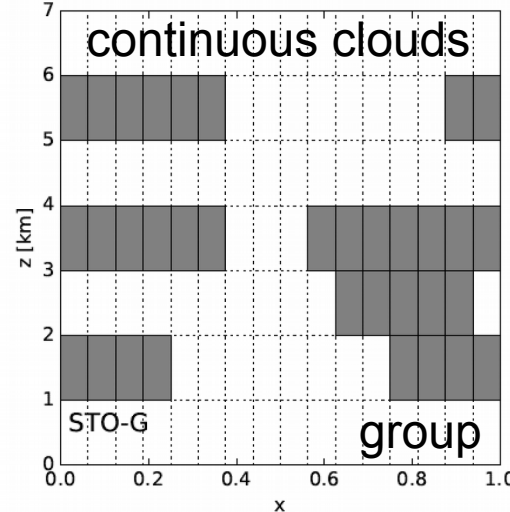
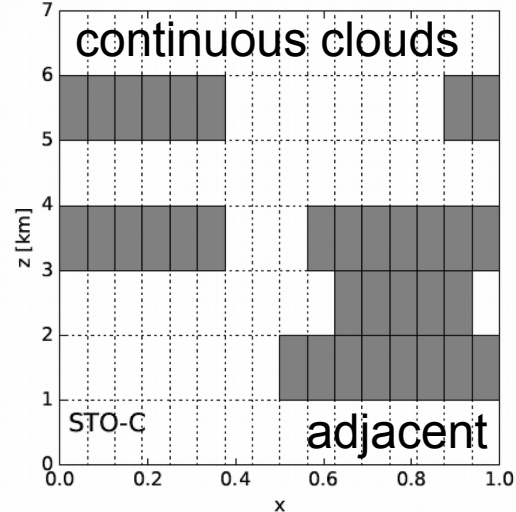
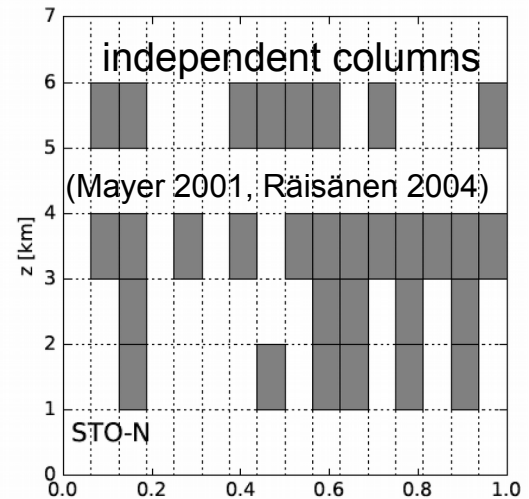
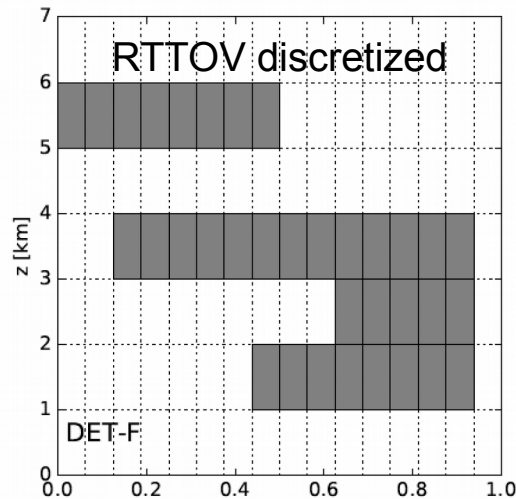
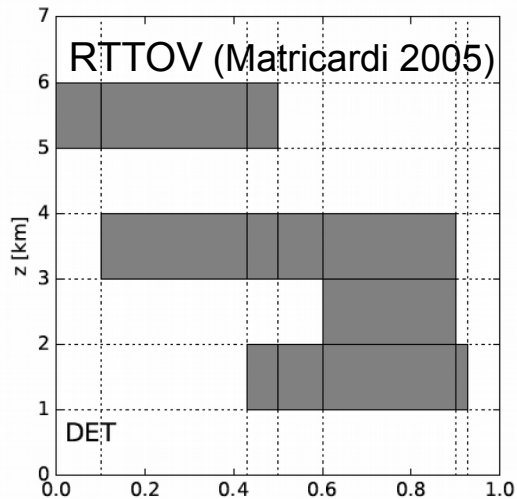
Stochastic schemes: Compute reflectance for one random realization
(spread quantifies uncertainty in cloud distribution)

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?

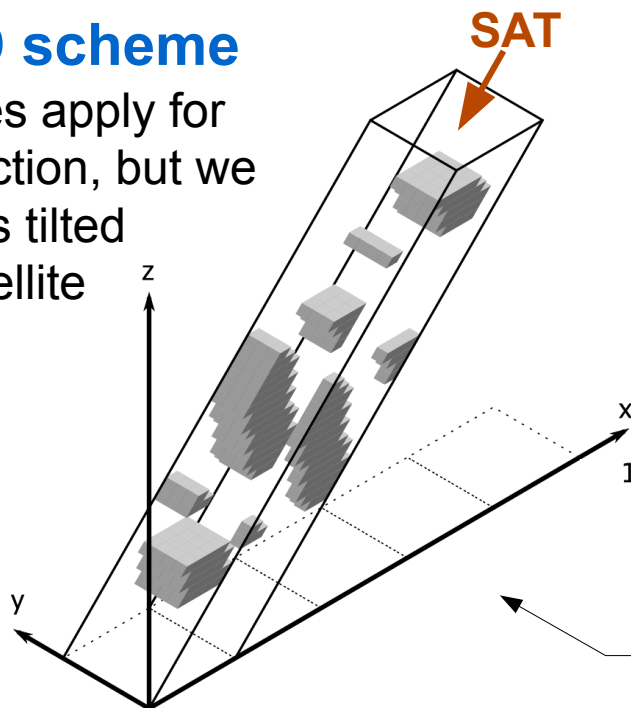
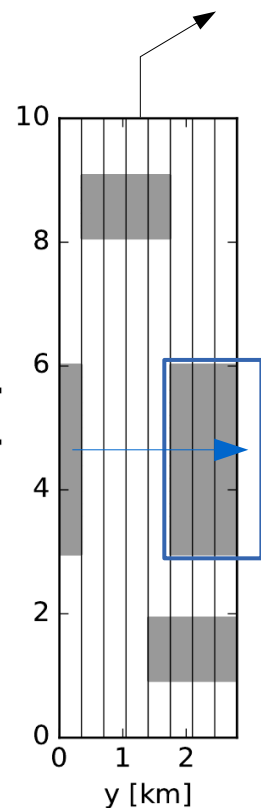


Common strategy: Subdivide column, fill subgrid cells according to overlap rules (different cloud size dist. possible), perform RT for each subcolumn, average results



A new 3D scheme

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

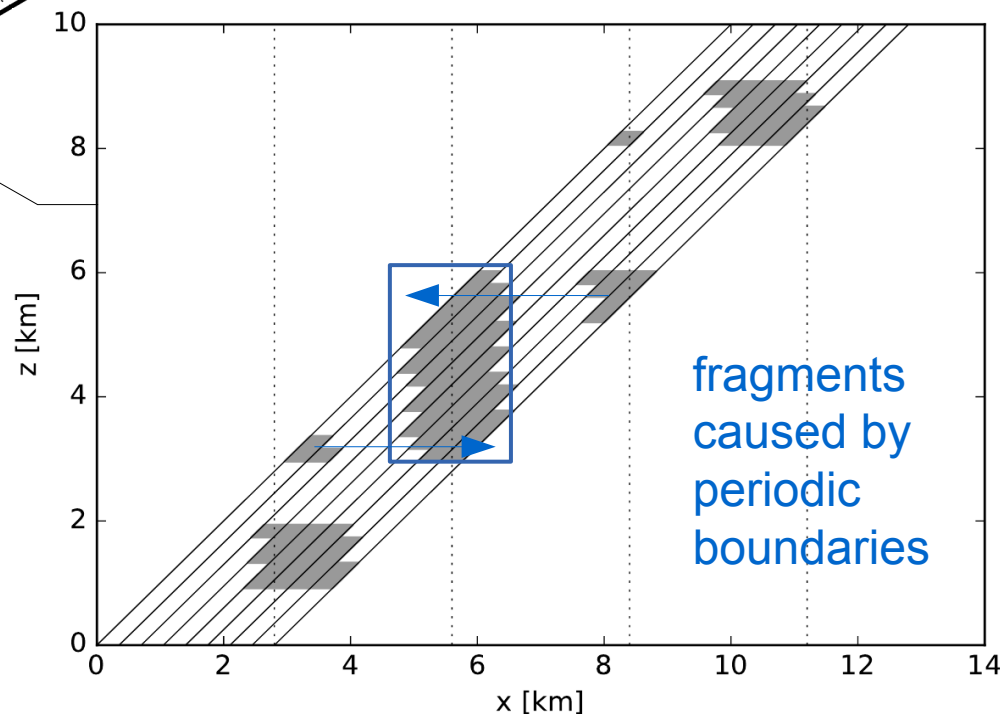


→ **Increased total cloud cover**
(also cloud sides contribute)

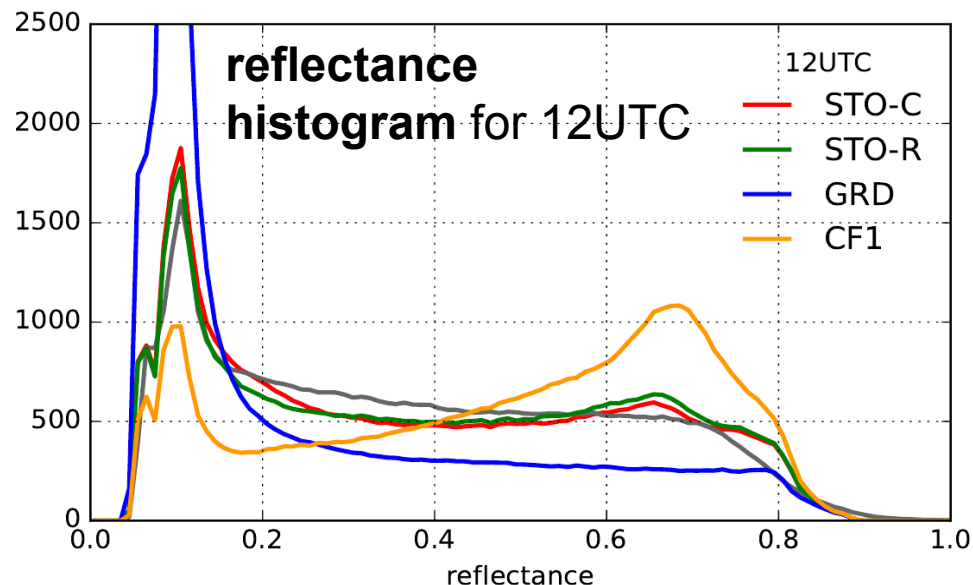
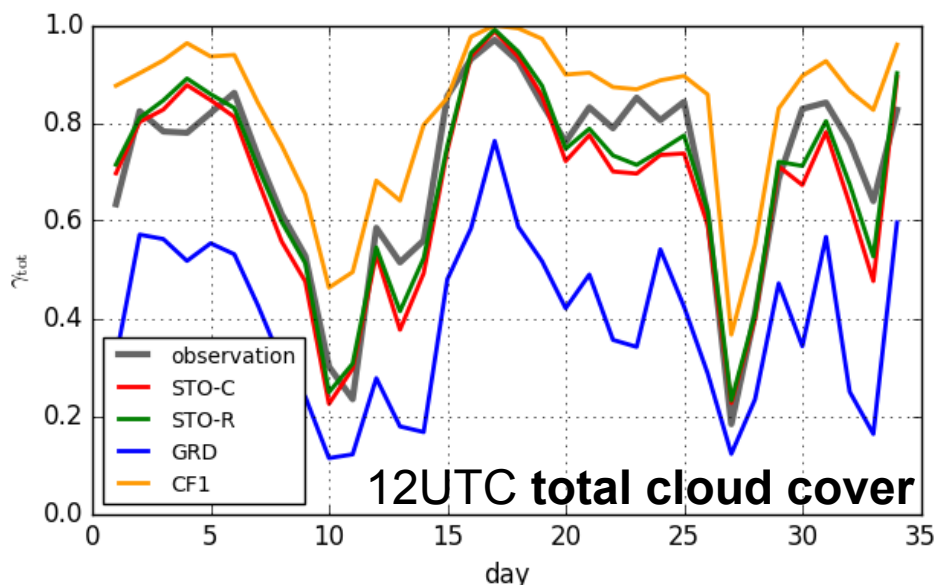
Not more expensive than 2D schemes.

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

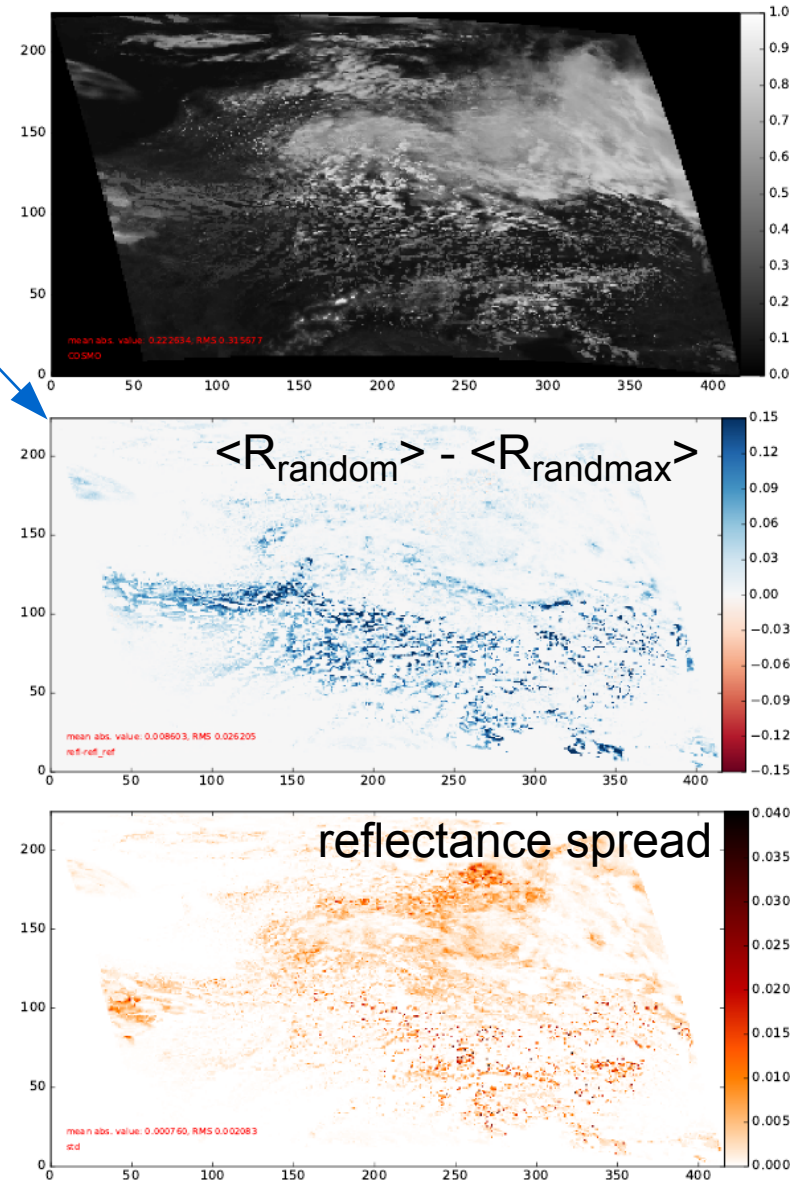
Mean reflectances of 2D stochastic schemes are very similar, also to deterministic schemes. ~ 10 subcolumns are sufficient.

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

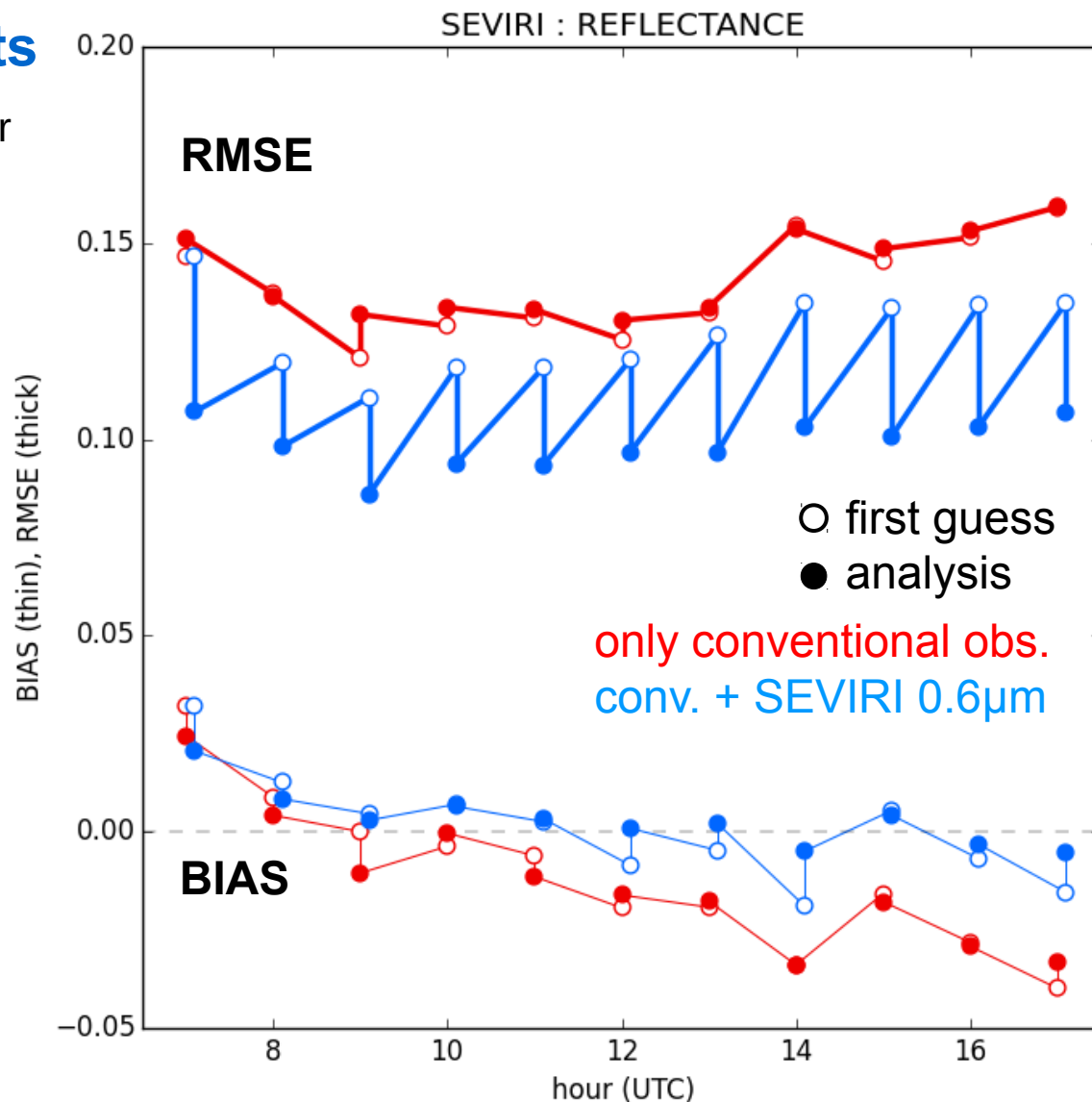
Setup:

40 member LETKF
1h assimilation interval
0.6 μ 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 an 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...

