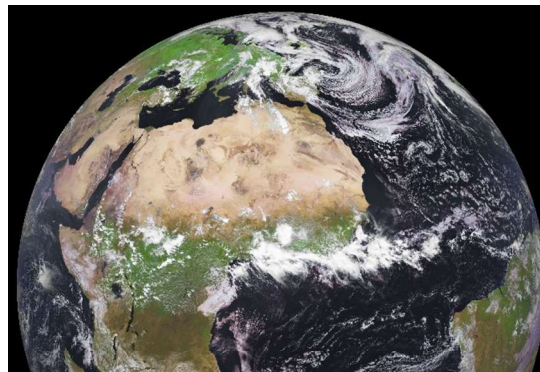


All-sky DA of visible channels of SEVIRI in ICON-D2-KENDA

COSMO General Meeting, KENDAScope Session

Liselotte Bach & Thomas Deppisch



*with C. Stumpf, L. Scheck
R. Faulwetter, M. Bender
A. Schomburg
C. Köpken-Watts, C. Schraff
A. de Lozar*

Observations

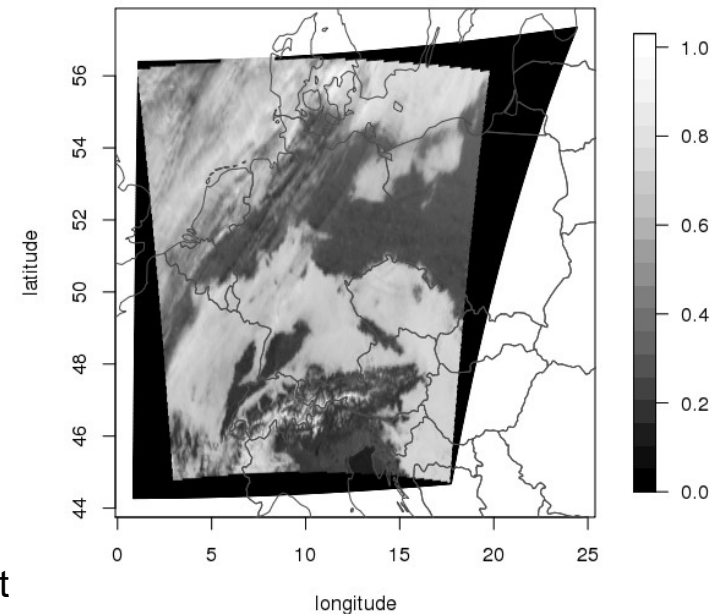
- Imager channel in the visible spectral range ($0.6 \mu\text{m}$)
- SEVIRI instrument on geostationary MSG ($0^\circ/0^\circ$)
 - Horizontal resolution: $6 \text{ km} \times 3 \text{ km}$ (Central Europe)

What is reflectance?

- Percentage of infalling solar radiation that is reflected by clouds and the earth's surface

Important characteristics

- Availability limited to day time
- Also sensitive to snow (alps!), volcanic ash, Saharan dust



All-sky data assimilation

Clouds reside in meteorologically interesting regions

- Tropical cyclones, fronts, convection, low stratus
- Clouds cover roughly 67 % of the earth's surface

Traditional all-sky satellite data assimilation

- Gain of vast amounts of temperature- and humidity-sensitive satellite data

But what is our goal?

- Assimilate visible channels directly sensitive to
 - Cloud water mass
 - Cloud optical properties
 - Cloud positions
 - Water vapour
 - Surface albedo



Why visible satellite data?

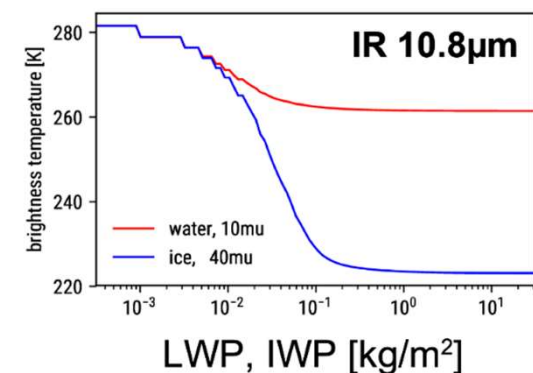
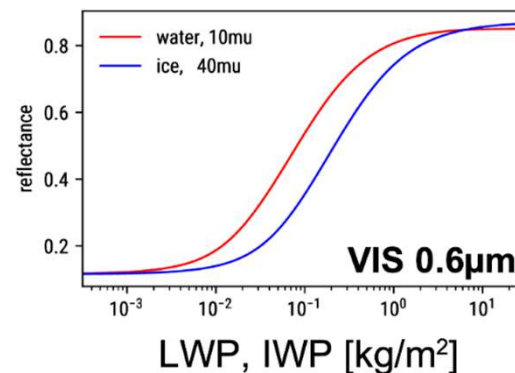
How does VIS differ from IR?

- Sensitive to cloud properties (VIS) rather than to temperature-humidity mixture (IR)
- Sees *also* boundary layer clouds (convective initiation, low stratus)
- Sensitive to a much larger range of LWP / IWP than IR
- Except for very small LWP / IWP (thin cirrus)

Which forecast impact can we expect?

- Cloud positions ~ precipitation
- Cloud optical depth ~ solar radiation
- Processes related to solar radiation, e.g. surface fluxes

1. Warnings high impact weather
2. Solar power forecasting
3. Flight meteorology (visibility)



Data assimilation methodology

Simulation of visible satellite pictures

RTTOV-MFASIS forward operator

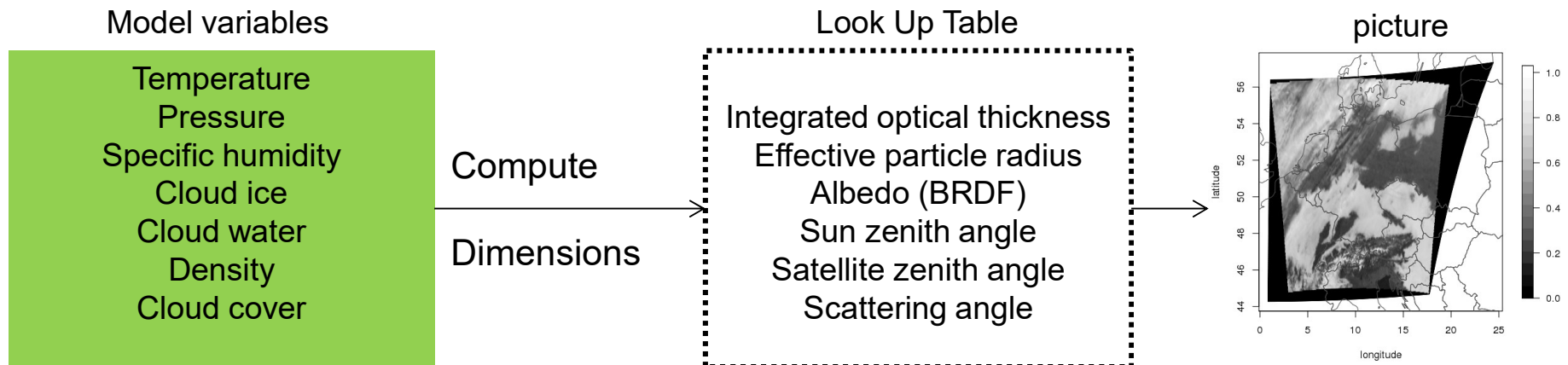
- ➔ Fast & accurate radiative transfer method MFASIS (Scheck, 2016)
- ➔ DA in operations conceivable for the first time
- ➔ Look-Up Table Approach; vertical integrals instead of vertical distribution
- ➔ Treatment of cloud variables: avoid interpolation (nearest neighbor)
- ➔ Ongoing developments: NIR-channel, aerosol, neural networks

Micro- and macro-physical assumptions

Cloud properties based on parameterization of effective radii (Reff):

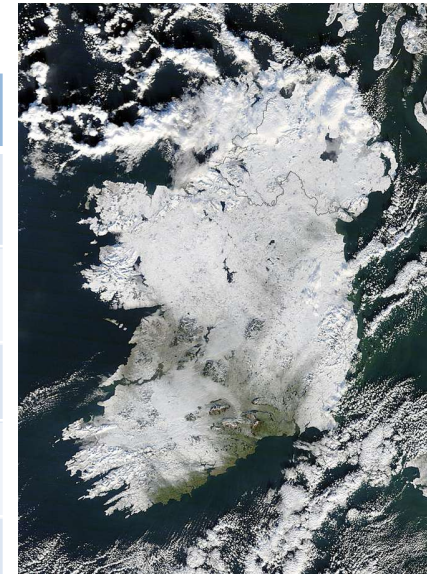
- Deff-scheme: Martin 1994 (cloud water)
- Baum-scheme: Mc Farquhar 2003 (cloud ice)
- Or use of Reff from ICON

Cloud overlap: maximum random overlap, no horizontal inhomogeneity

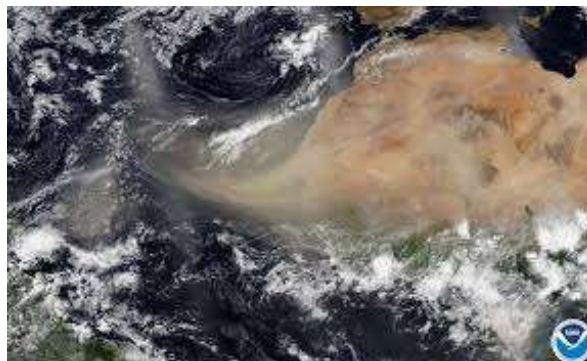


Quality control

Parameter	Reason for rejection
Sun zenith angle ($> 75^\circ$)	Missing 3D-effects in MFASIS Night
Model orography > 1100 m Cloud mask = „SNOW“	Misinterpretation of snow as clouds
Obs > 1.5	Missing 3D-effects in MFASIS
Saharan dust / volcanic ash Cloud mask = „DUST“	Misinterpretation as clouds
MFASIS operator flags	Magnitude of effective radius



Mixture of snow and clouds



Saharan dust outbreak

NWC-SAF cloud mask

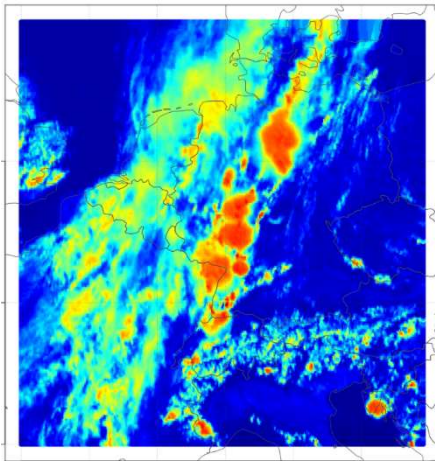
Use flags for snow, aerosol, volcanic ash as part of satellite preprocessing QC (satpp)

Full satellite picture to observations

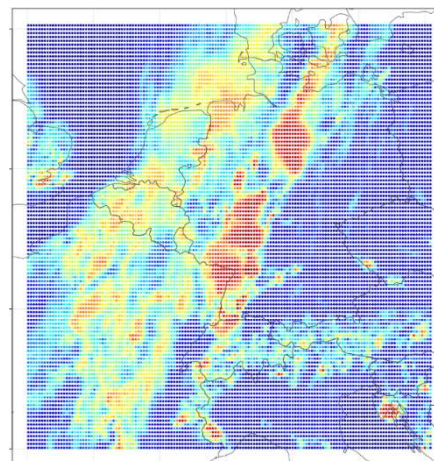
Data reduction (superobbing) to

- Balance remote sensing data and conventional observations
- Reduce representativity error and double penalty problems
- Account for assumption of spatially uncorrelated observation error
- Applied to both y and $H(x)$ after nearest neighbor interpolation of model columns to satellite grid

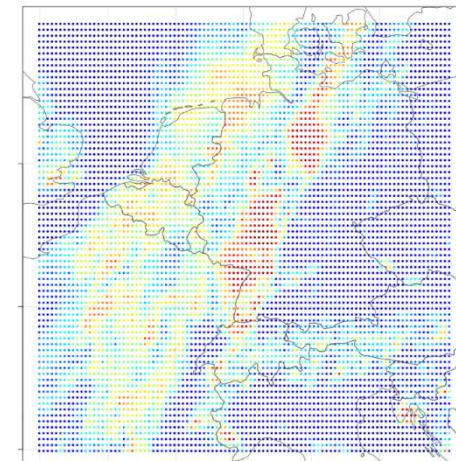
Full resolution



~12 km



~18 km



Vertical localization

Problem

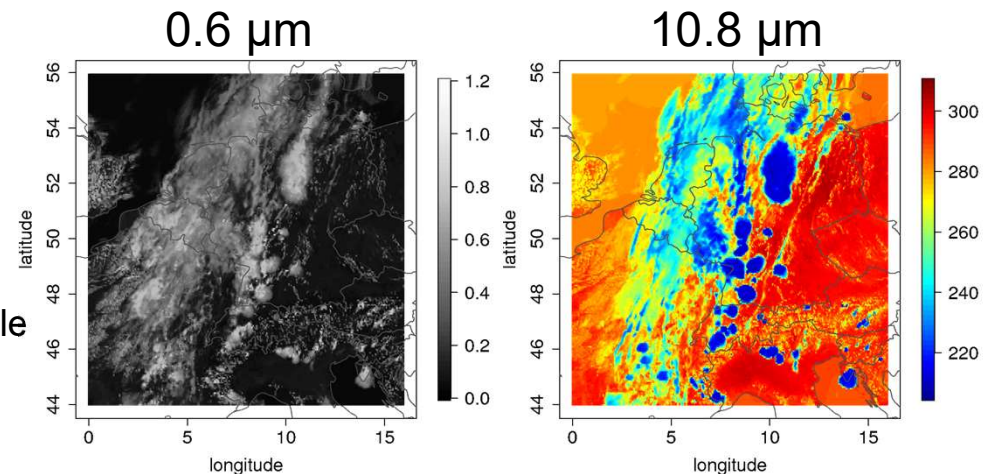
- Reflectance represents vertical integral over model column
- No height information, no information about vertical extent of clouds
- Attempts with vertical localization based on cloud products (cloud type, NWC-SAF)

How do we deal with that?

- Currently no vertical localization of vertical intervariable correlations
- Future: visible + infrared channels

Does it work?

- Mainly directly cloud-dependent processes are improved
- By tendency little impact through intervariable correlations (e.g. $\rho(T, REFL)$)



What can be possible predictors of obs error?

- Nonlinearity
- Sun zenith angle
- First guess departure ~ displacement error

Inflate obs error depending on first guess departure

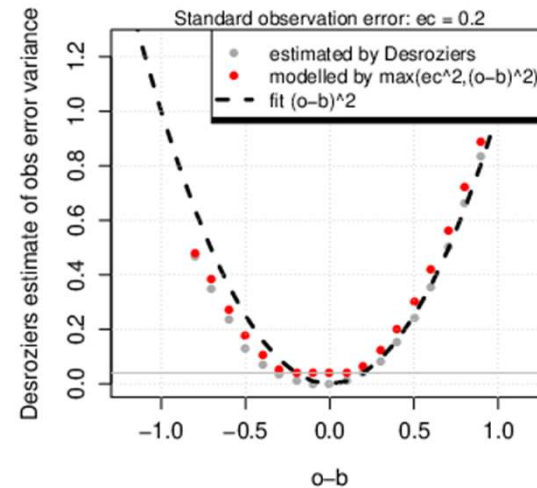
→ $\sigma_o^2 = \max(e_c^2, (o - b)^2)$

Account for spread

→ $\sigma_o^2 = \max(e_c^2, (o - b)^2 - \sigma_b^2)$

$Var(o - b) = Var(o) + Var(b) - 2 * Cov(o, b)$

Desroziers estimation stratified by first guess departure



$$\sigma_o = \frac{1}{N} \sum_{i=1}^N (o_i - a_i)(o_i - b_i)$$

Calibration of observed satellite pictures

Pure observation bias

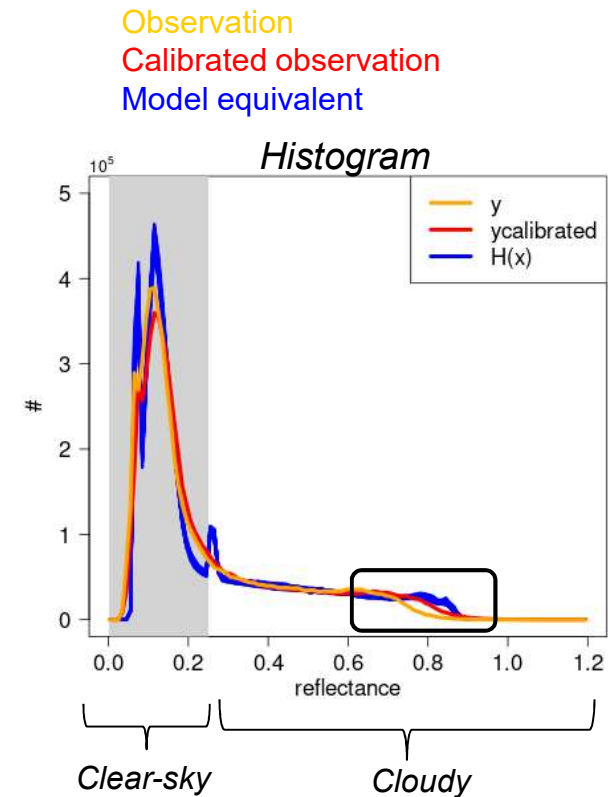
- SEVIRI visible channels ~ 8-10% too dark compared to the moon, MODIS satellite data (EUMETSAT)

How do we fix that problem?

- Calibrate satellite observations by fixed factor of 1.08
- Better agreement of histograms

What about remaining bias of first guess departures?

- Conditional bias in cloudy part of the histogram leads to detrimental forecast impact
- New histogram-based bias correction to stabilize impact



How to do bias correction?

What is our goal?

- Apply locally conservative corrections to every pixel of the simulated satellite picture such that histogram error vs. observed reflectances is reduced

Methodology

- Correction function: polynomial in reflectance weighted by sun zenith angle

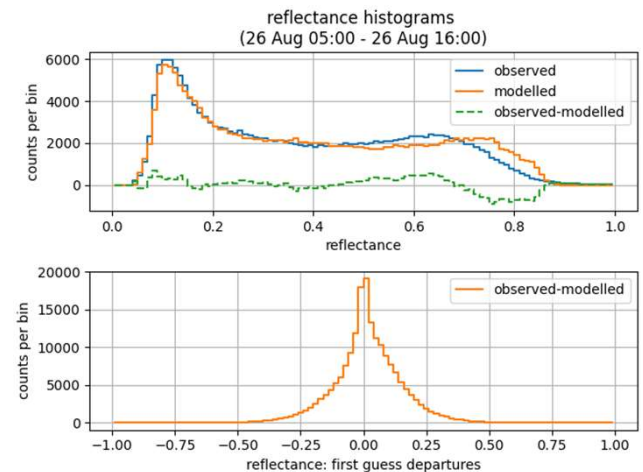
$$f(r, \theta) = \sum_{ij} c_{ij} T_i(r) U_j(\theta)$$

- Estimate coefficients c bei minimizing the following cost function

$$J[c^a] = \alpha \underbrace{\|c^a - c^b\|_{B_1}^2}_{\text{Prior}} + \beta \underbrace{\|c^a - c^0\|_{B_2}^2}_{\text{Relaxation}} + \|hist(fg) - hist(obs)\|_{R^{-1}}^2 \text{ (Kalman Update)}$$

Issues

- How adaptive should the bias correction be?
- How strongly does reflectance histogram bias vary depending on diurnal cycle / weather regime?

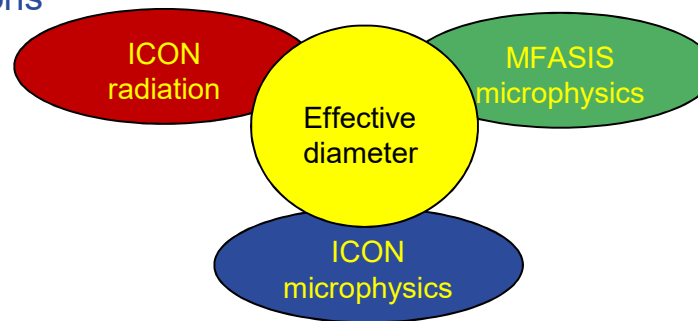


What else can we do concerning bias?

1. Statistical bias correction in data assimilation
 - Bias of first guess departures
2. Tune ICON model vs. satellite observations
 - Better frequency distributions
 - Reduced compensating error

Inconsistent microphysical assumptions

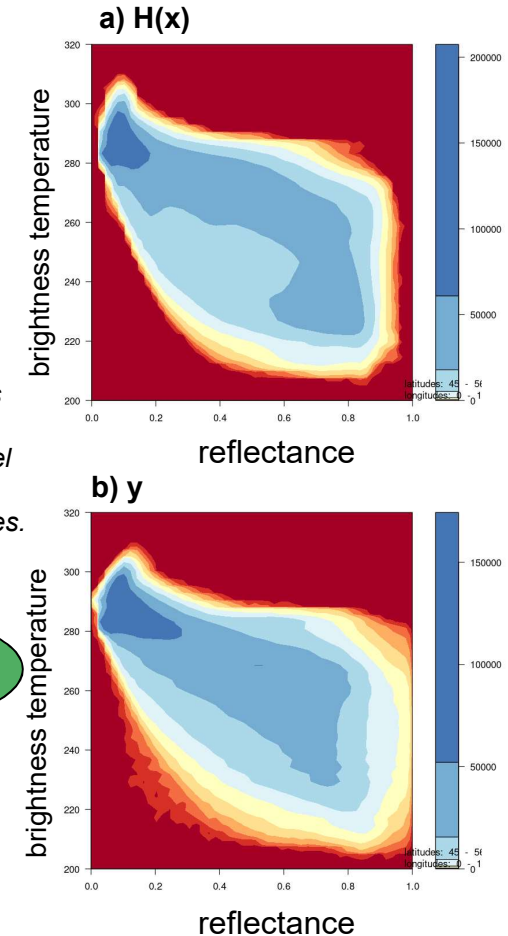
- Horizontal heterogeneity
- Vertical cloud overlap
- Effective diameters



How do we deal with that?

- Effective diameter parameterized (1MOM) / forecasted (2MOM) by ICON microphysics
- Used in ICON microphysics, ICON radiation, MFASIS forward operator

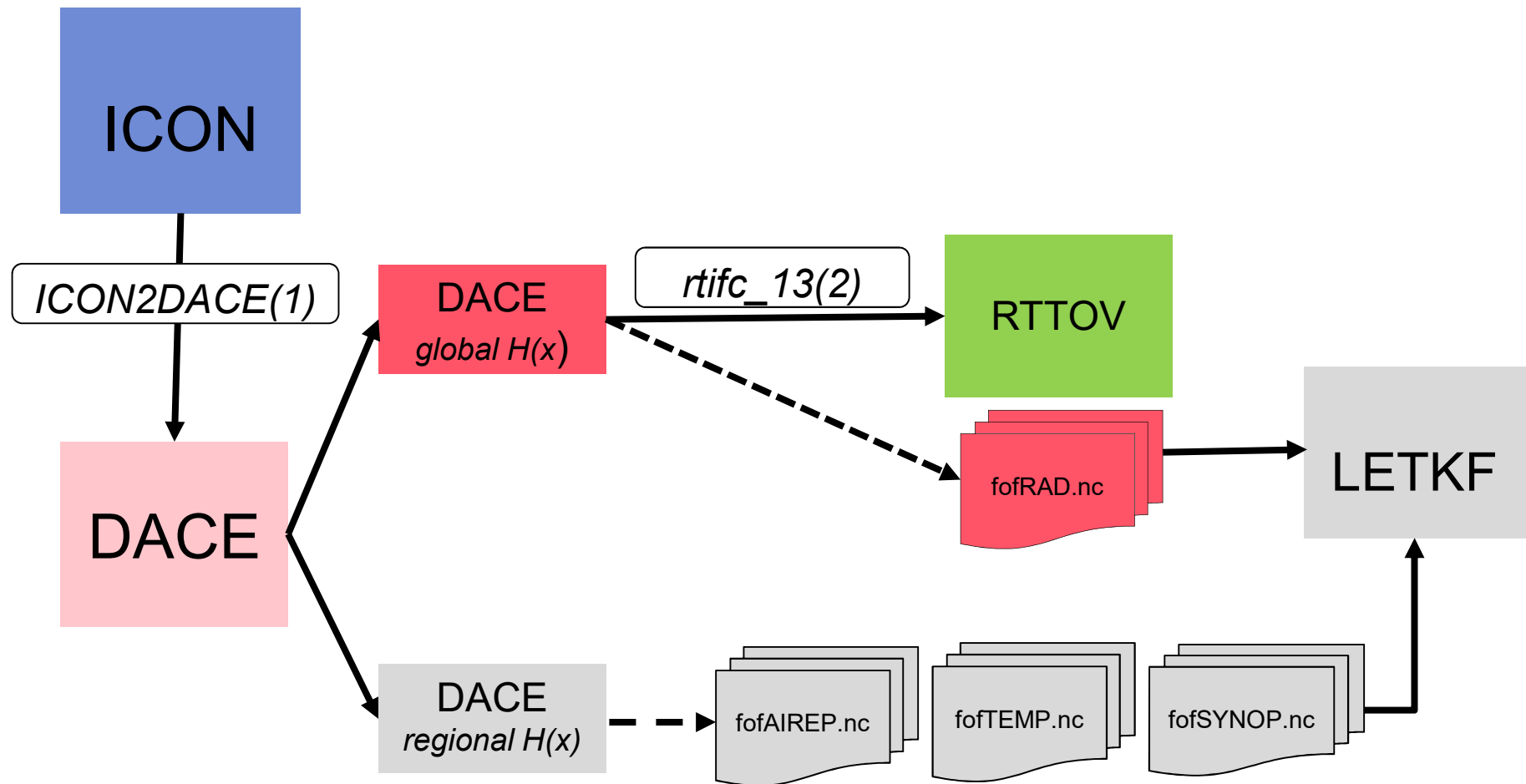
Figure shows joint frequency distributions of reflectance ($0.6 \mu\text{m}$) and the infrared window channel ($10.8 \mu\text{m}$) for a) simulated, b) observed satellite pictures.



NWP framework

Program structure

1. Interface from ICON to DACE that calls forward operators during model run
2. Interface from dace to rtov



Impact experiments with ICON-D2

Experiments

General Settings

- ICON-D2, cpcv-bugfix, 1MOM
- *Offline interface to RTTOV-MFASIS*
- *2.August 2020 – 26.August 2020*
- *DA of 0.6 μm (LB)*
- *DA of water vapour channels (AS)*

Observations

- 1 satellite picture / hour [*@60min*]
- Conv. obs AIREP, TEMP, SYNOP, MODES
- Latent Heat Nudging
- 3D radar reflectivities + radial winds
- Calibration of observations 1.08

DA settings

- 12 km superobbing scale
- 35 km horizontal localization
- Observation error 0.2
- No vertical localization
- No bias correction

Observations are rejected if

- Sun zenith angle $> 75^\circ$ (3D-effects)
- Model orography > 1100 m (snow)
- Obs > 1.5 (missing 3D-effects)
- Boundary of domain
- *Saharan dust, snow, nonlinearities*

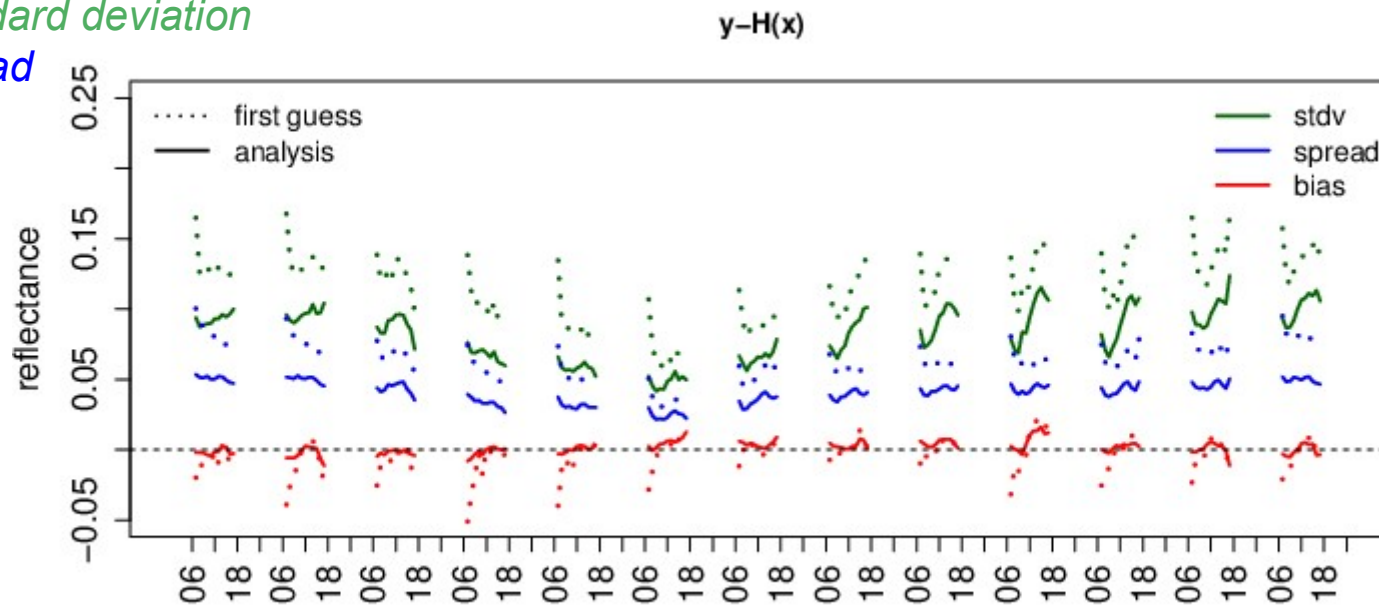
Results in DA cycle

Reflectance statistics in DA cycle

Standard deviation

Spread

Bias

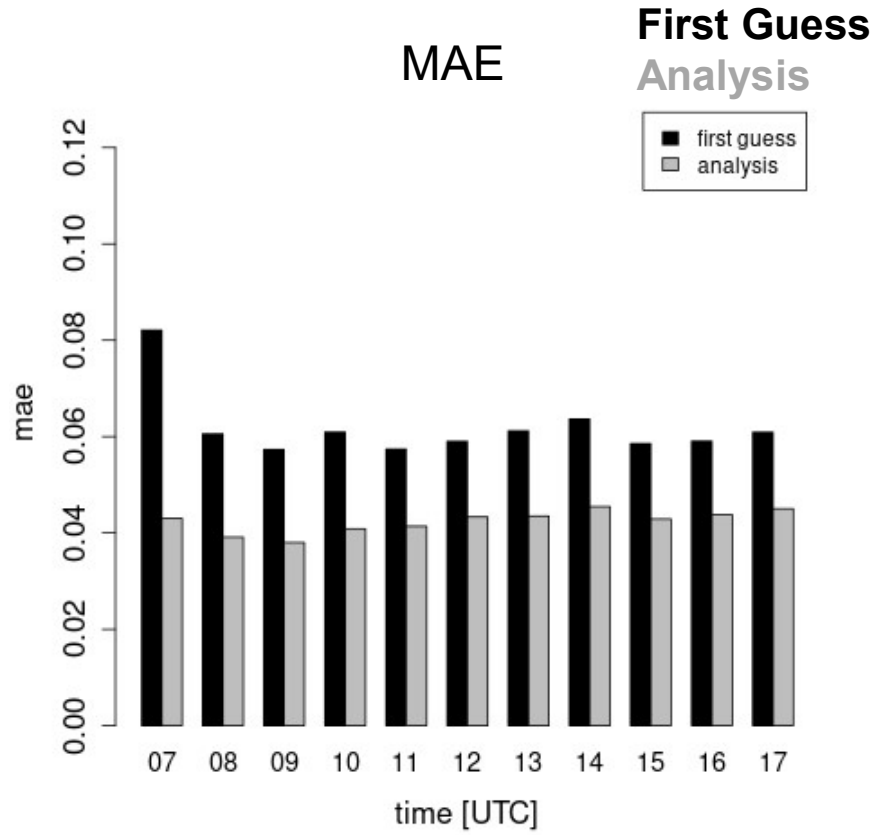
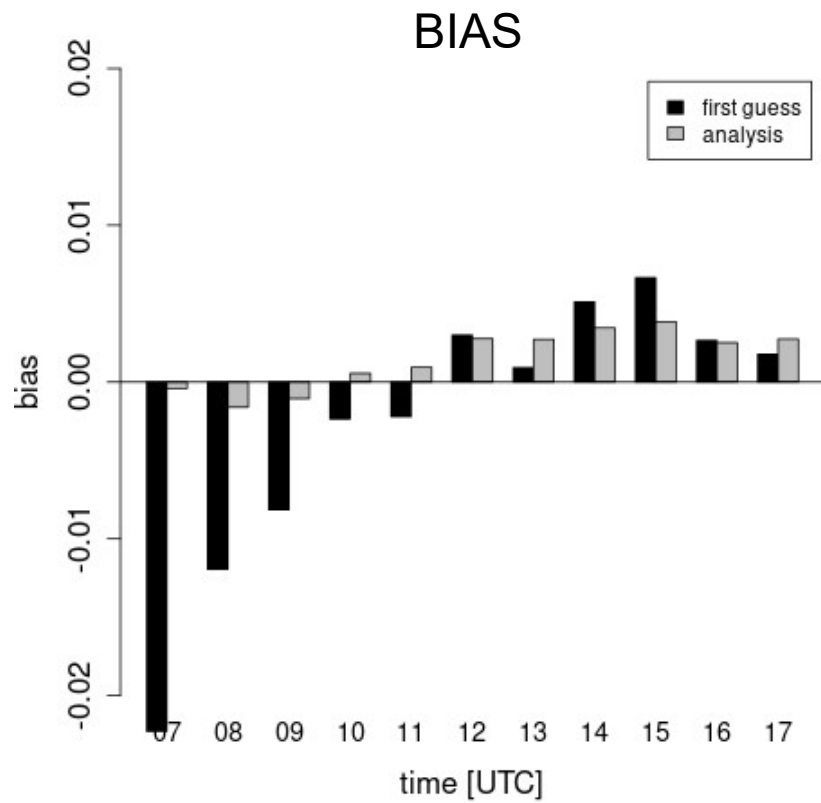


On superobbing scale



DA allows to better understand model error

Error reduction in reflectance in DA cycle



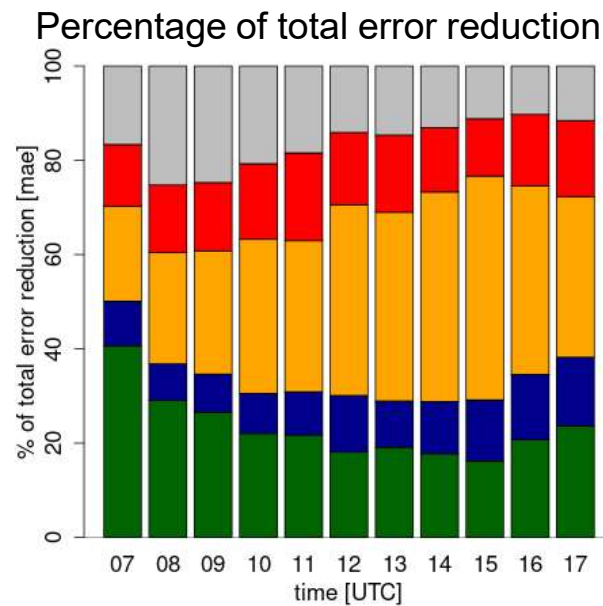
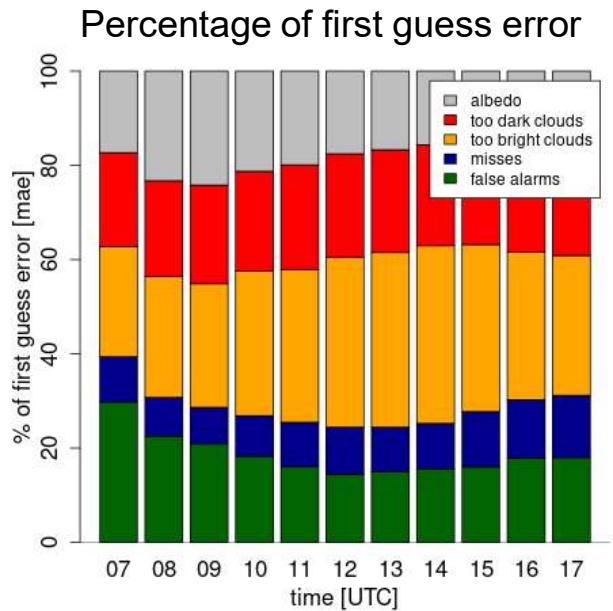
Threshold : REFL=0.25

REFL < 0.25 : Clear-sky

REFI > 0.25 : Cloudy

DA allows to better understand model error

Error reduction in reflectance categories



Clear sky error
Cloud too dark
Cloud too bright
Cloud is missing
Cloud is false alarm

Ambiguous reasons for errors

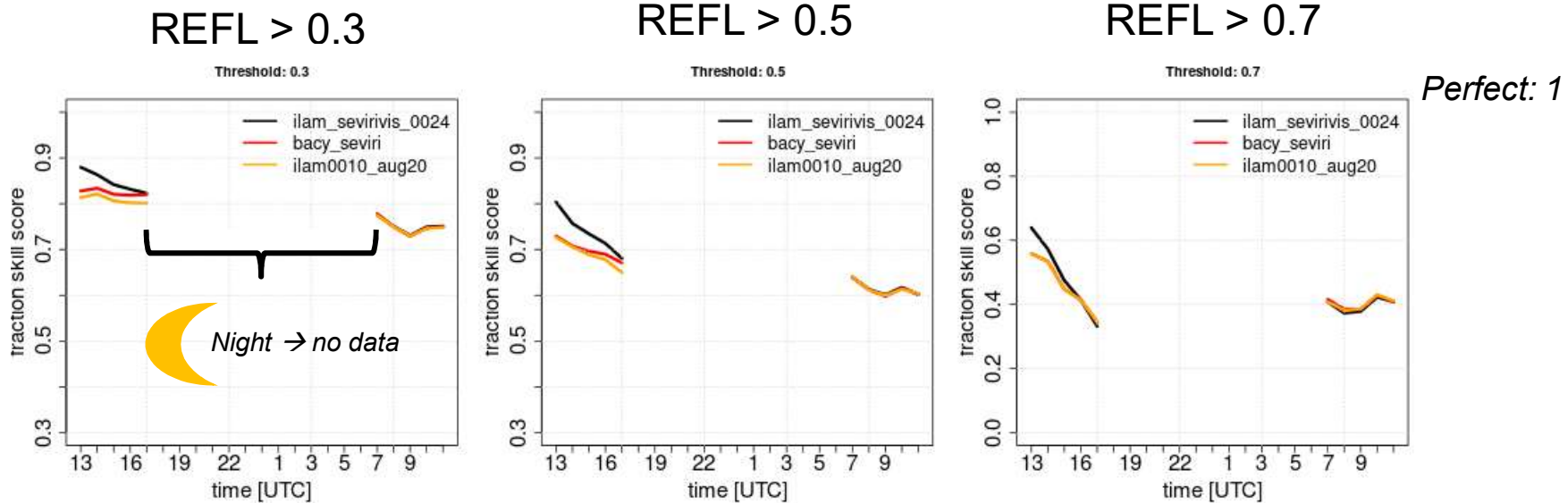
- Clear Sky : Erroneous BRDF-climatology, missing aerosol in LUT, water vapour erroneous
- Too dark : Missing 3D effects, too little water mass, wrong water phase (ice), too big particles, too few particles
- Too bright : Too much water mass, too small particles, wrong water phase (water), too many particles
- Cloud is missing : Cloud position error, model cloud is missing, threshold error
- False alarm cloud : Cloud position error, cloud number overestimation, threshold error



Forecast impact

Reflectance (Fraction skill score)

FSS, 7 satellite pixels
 Initialized at 12 UTC
 24 days

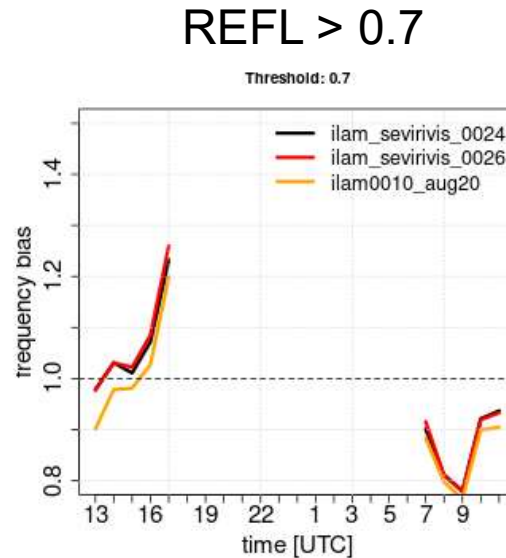
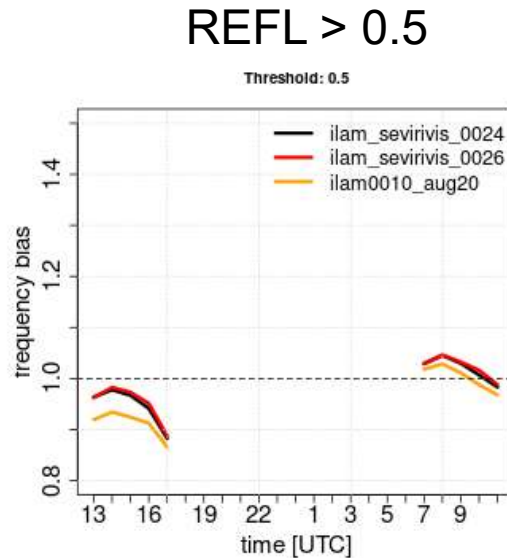
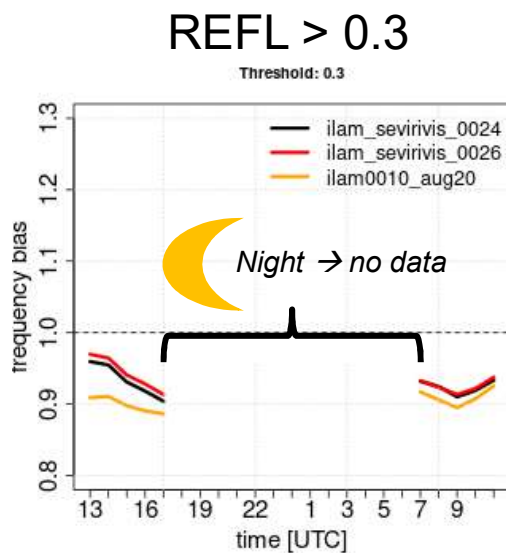


REFL > 0.3 : all clouds
 REFL > 0.5 : optically medium thick and thick clouds
 REFL < 0.7 : optically thick clouds



Reflectance (Frequency bias)

FBI
 Initialized at 12 UTC
 24 days

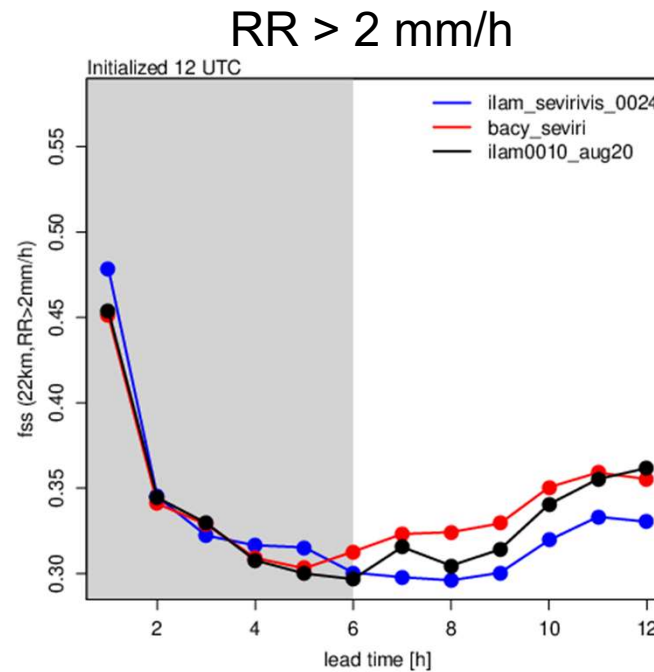
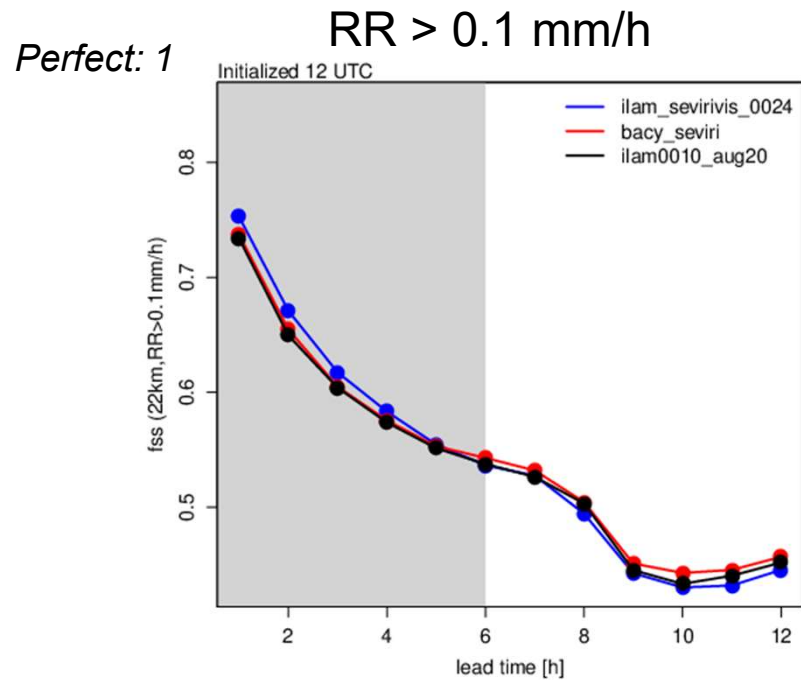


Perfect: 1

REFL > 0.3 : all clouds
 REFL > 0.5 : optically medium thick and thick clouds
 REFL < 0.7 : optically thick clouds

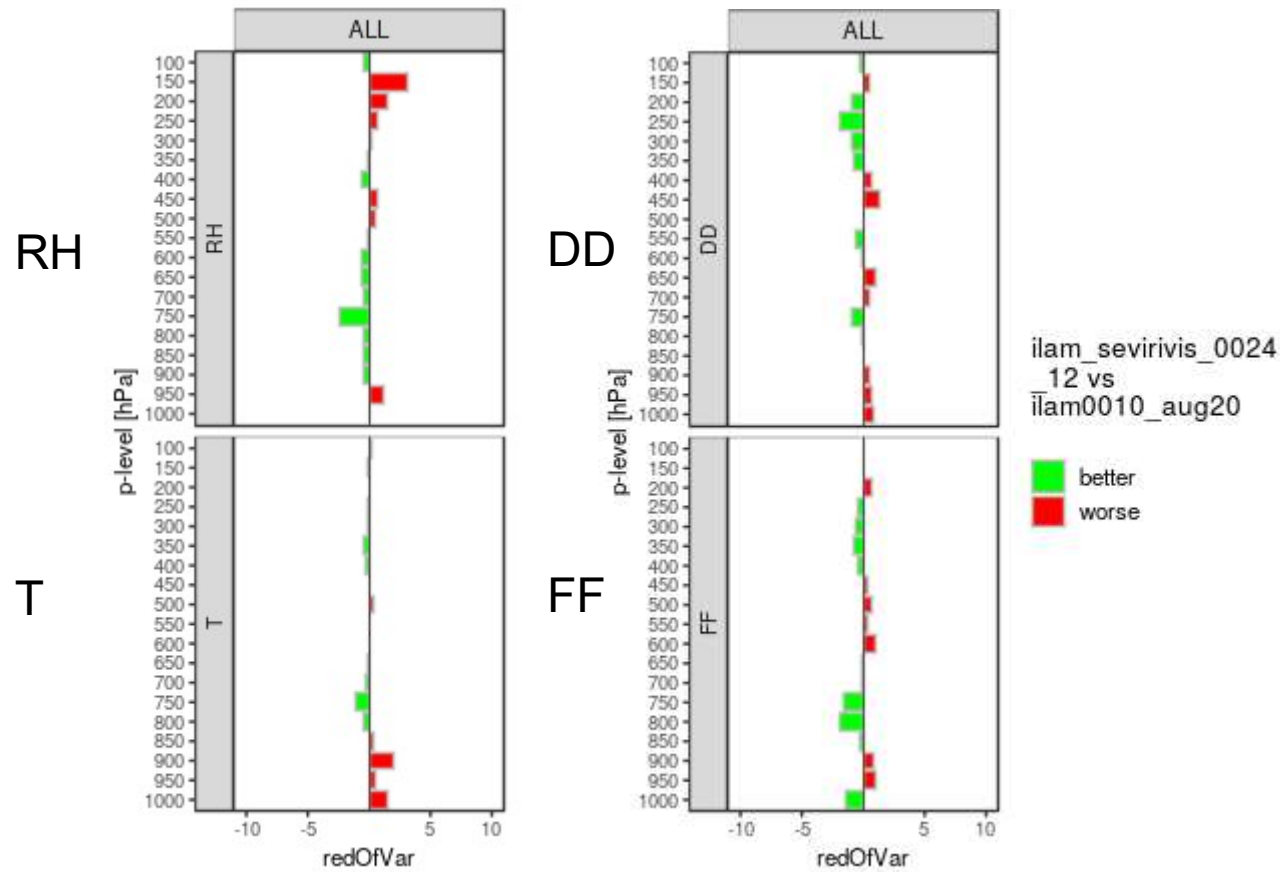


Precipitation



Upper Air Verification

CONV + RADAR + SEVIRI-VIS better
CONV + RADAR better



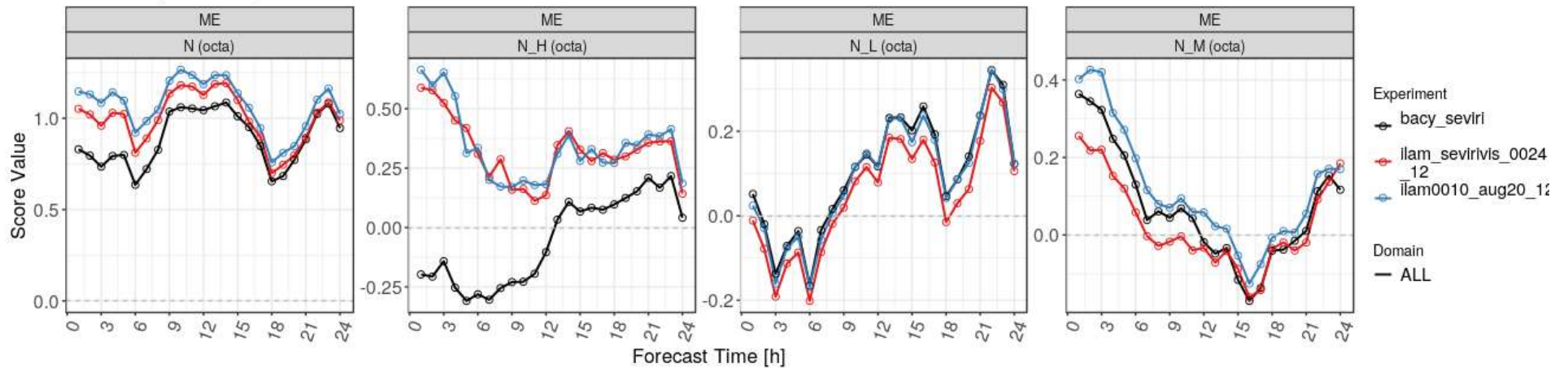
Initialized 12 UTC
24 days

CONV + RADAR + SEVIRI-WV CONV + RADAR + SEVIRI-VIS CONV + RADAR

Cloud cover

2020/08/03-12UTC - 2020/08/26-00UTC
INI: 12 UTC, DOM: ALL, STAT: ALL

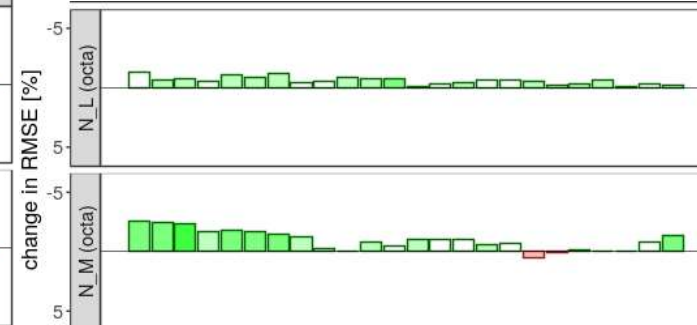
- ✓ Improvement of cloud cover bias over whole forecast horizon (24h)
- ✓ Error reduction in RMSE



Total



High clouds



Low clouds

Medium clouds

CONV + RADAR + SEVIRI-VIS better

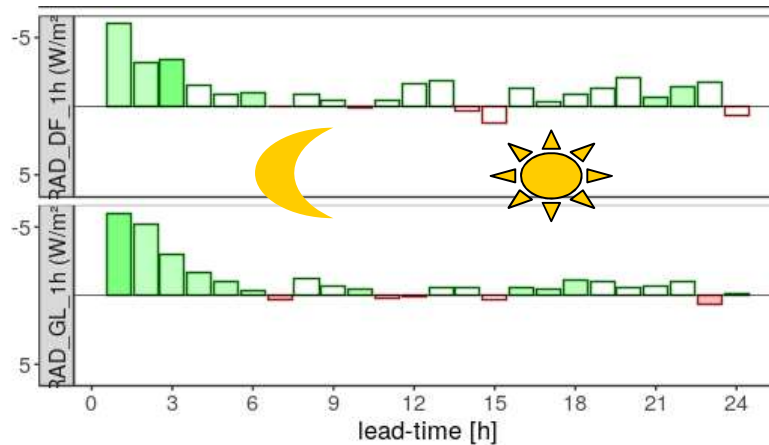


Verified vs. SYNOP observations

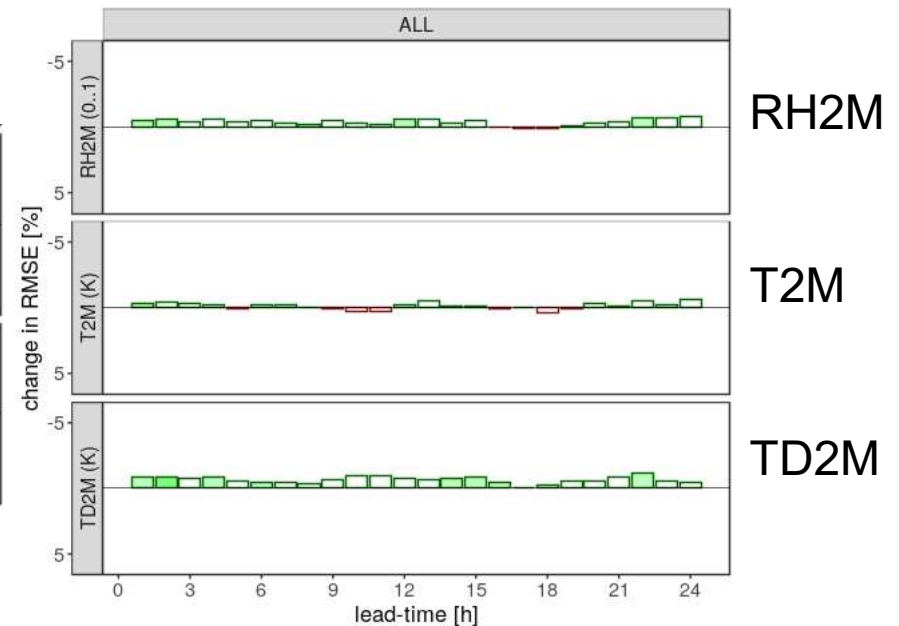
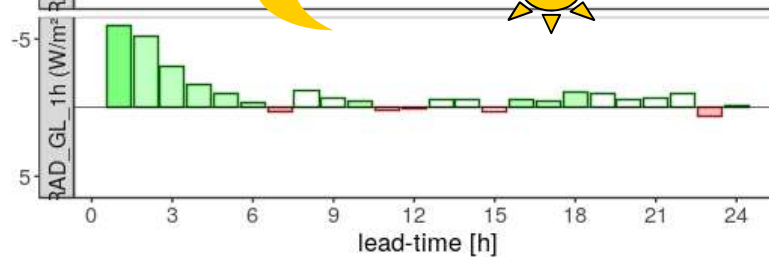
Short-wave radiation and surface variables

- ✓ Improvement of radiation through improved cloud cover / cloud optical depth
- ✓ Through better radiation better screen-level temperature and humidity

Diffuse radiation



Global radiation



CONV + RADAR + SEVIRI better

CONV + RADAR better



**The impact is very dependent on
the model version**

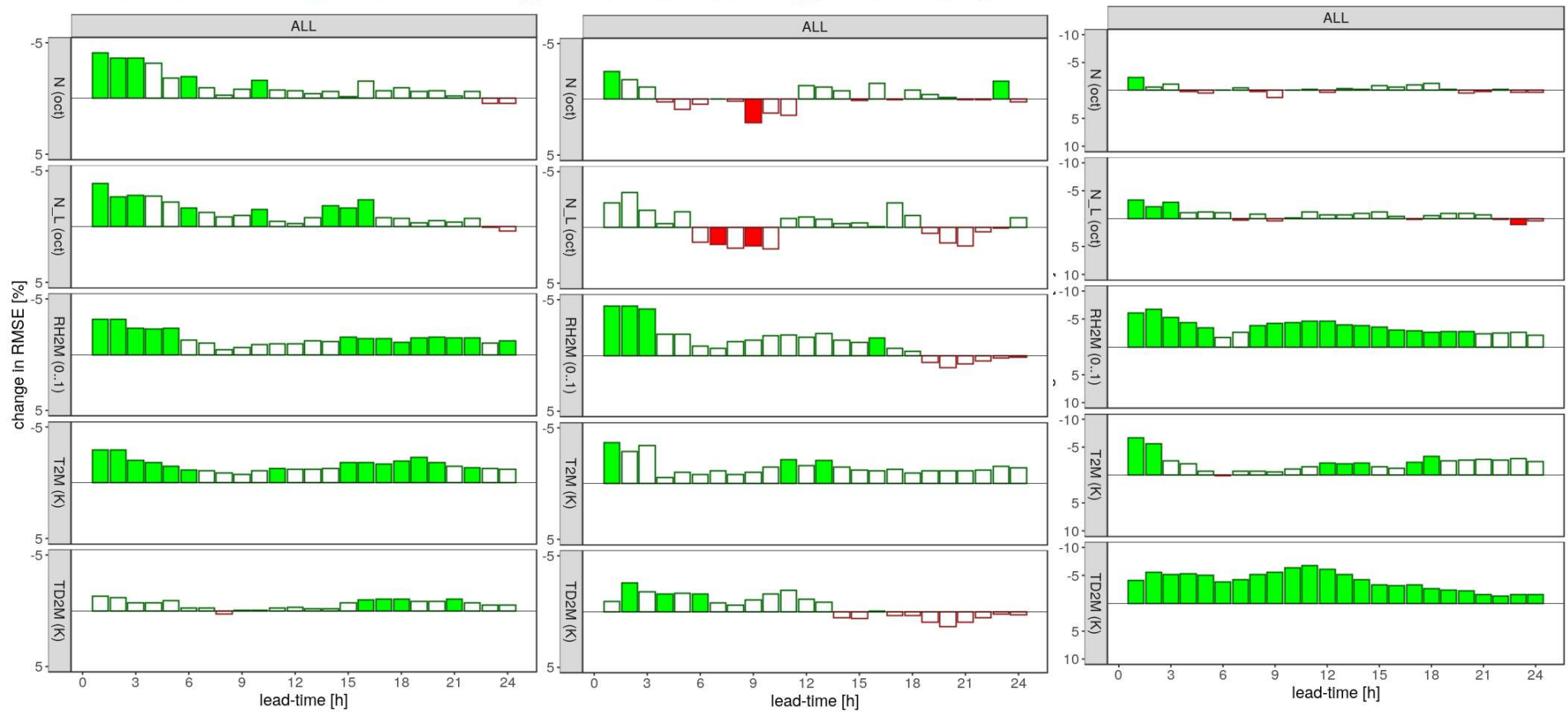


Dec 2018

June 2019

July 2019

5% Reduction of RMSE [%], INI; 12UTC, SIGTEST: TR
 5% Reduction of RMSE [%], INI; 12UTC, SIGTEST: TR
 10% Reduction of RMSE [%], INI; 12UTC, SIGTEST: TR



Conclusion

- SEVIRI-VIS shows to have impact on reflectance, precipitation, global radiation and surface variables
- Results are highly different between different model versions

- New experiments run with technically mature 4D-LETKF, i.e. RTTOV is called during ICON run
- Parallel-operations in SINFONY-RUC and ICON-D2 are pursued in the near future
- Preparations for data bank arrival times, satellite preprocessing (satpp), NUMEX finished

Final requirements

- Understanding differences in simulated satellite pictures (ICON, VISOP, offline interface)
- Final model tuning for 2-Moment-Scheme
- Working bias correction (for ICON-D2) and alert system
- Experiments