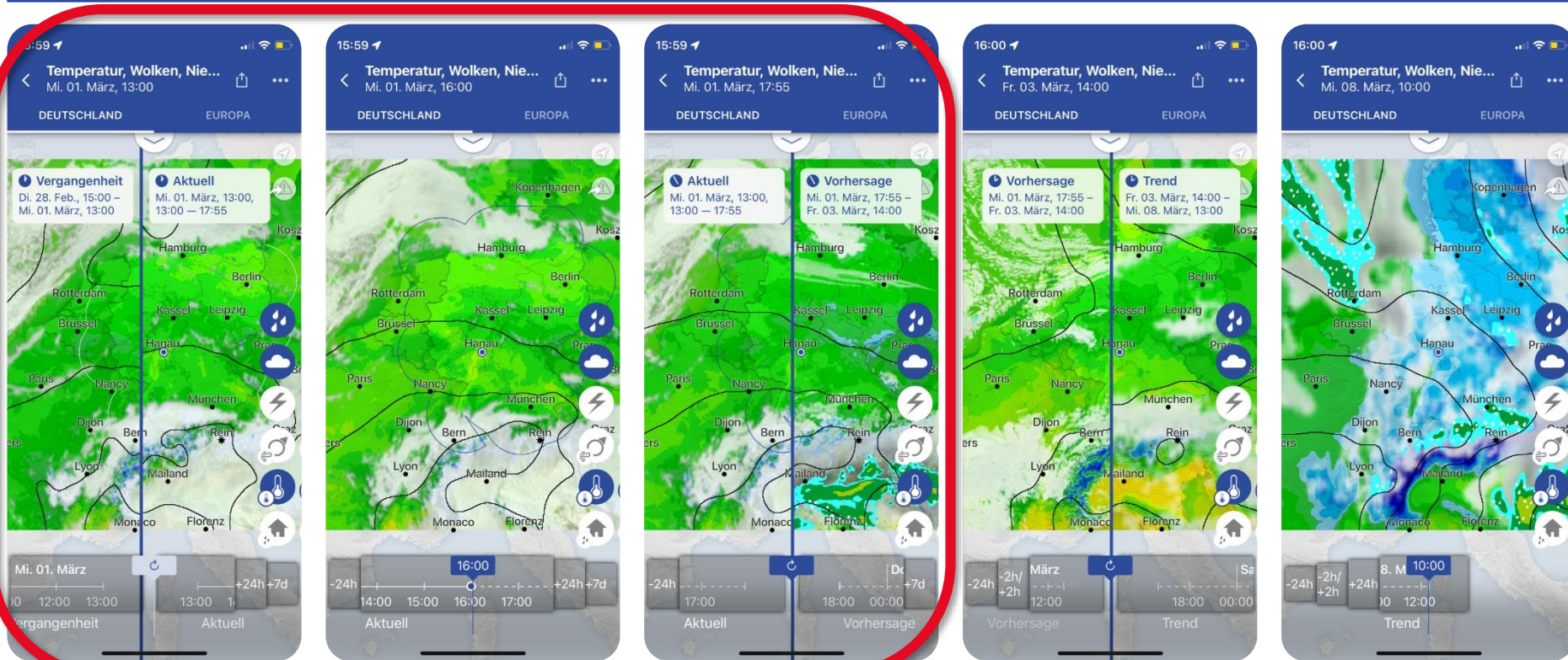


# Enhancing Numerical Weather Prediction with Opportunistic and Crowdsourced Observations: Insights from NetAtmo and Microwave Link Assimilation

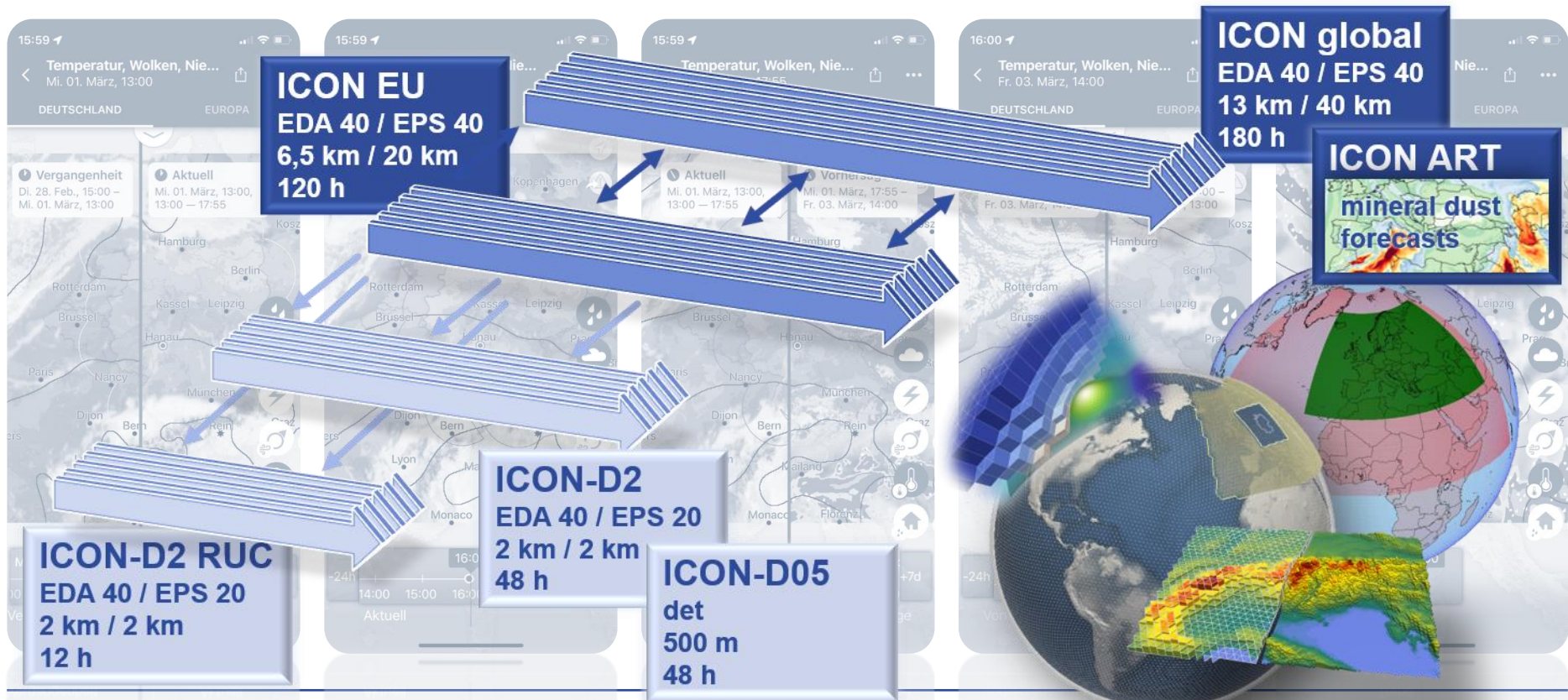
Stefanie Hollborn  
FE12 – Observation Modelling and Verification  
FE1 – Meteorological Analysis and Modelling



# Range of DWD Weather Forecasts



# NWP Model Chain





# NWP DA Methods

**ICON EU**  
EDA 40 / EPS 40  
6,5 km / 20 km  
120 h

**Regional System:  
4D-LETKF**

also available:  
particle filter method LMCPF

also available for  
conventional observations:  
cEnVAR  
EnVAR plus LETKF

in preparation:  
AIVAR

**Global System:  
EnVAR plus LETKF**

also available:  
particle filter method LMCPF

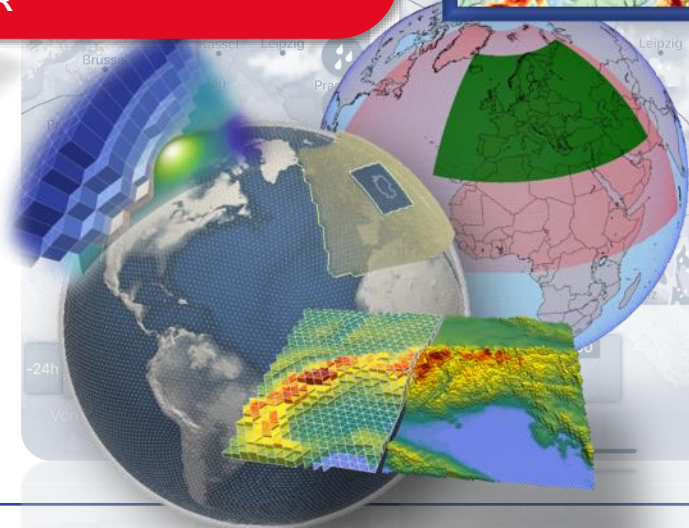
in preparation:  
FGAT and 4D-EnVAR  
AIVAR

**ICON ART**

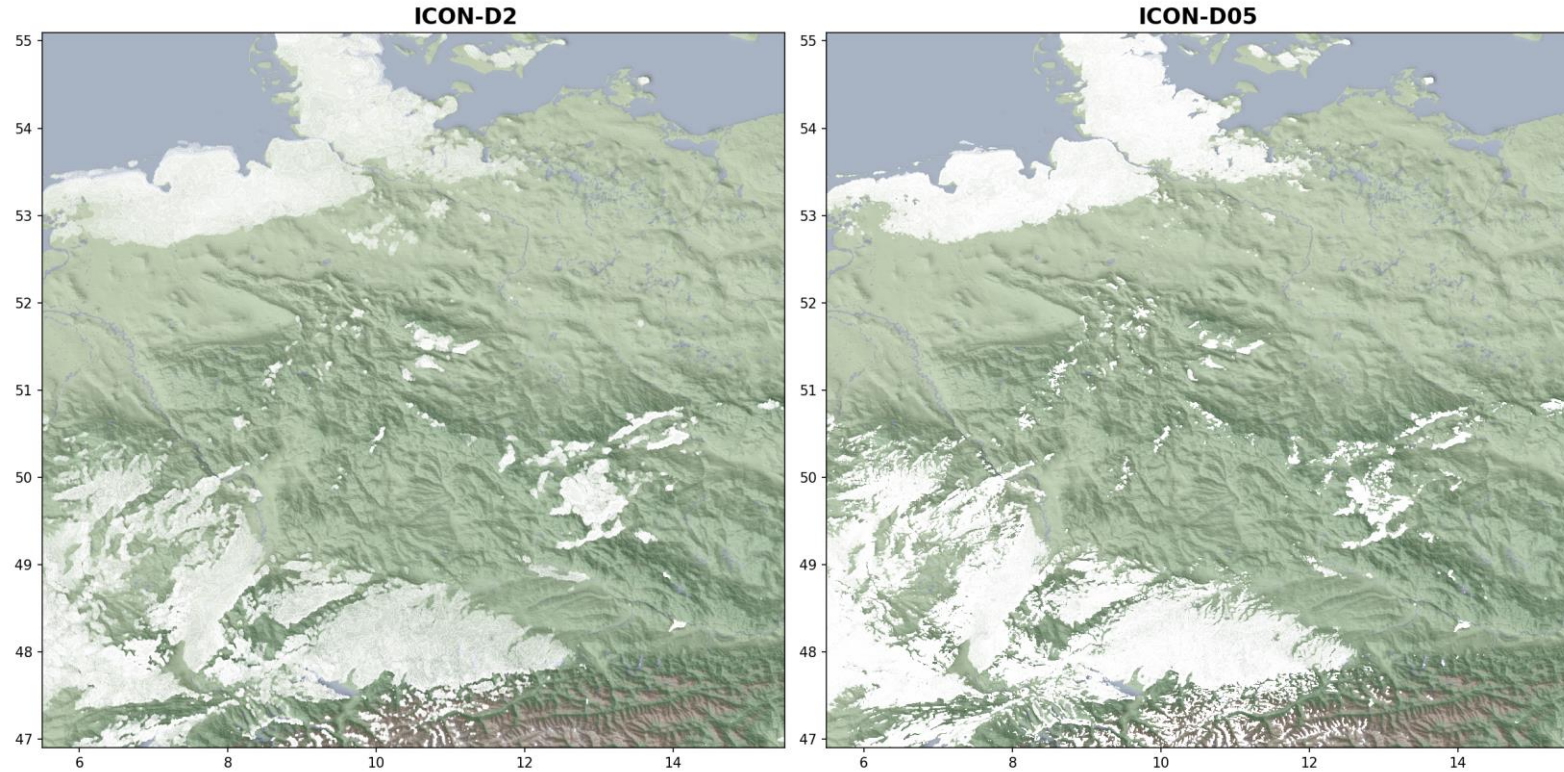
mineral dust  
forecasts

**ICON-D2**  
EDA 40 / EPS 20  
2 km / 2 km

**ICON-D05**  
det  
500 m  
48 h

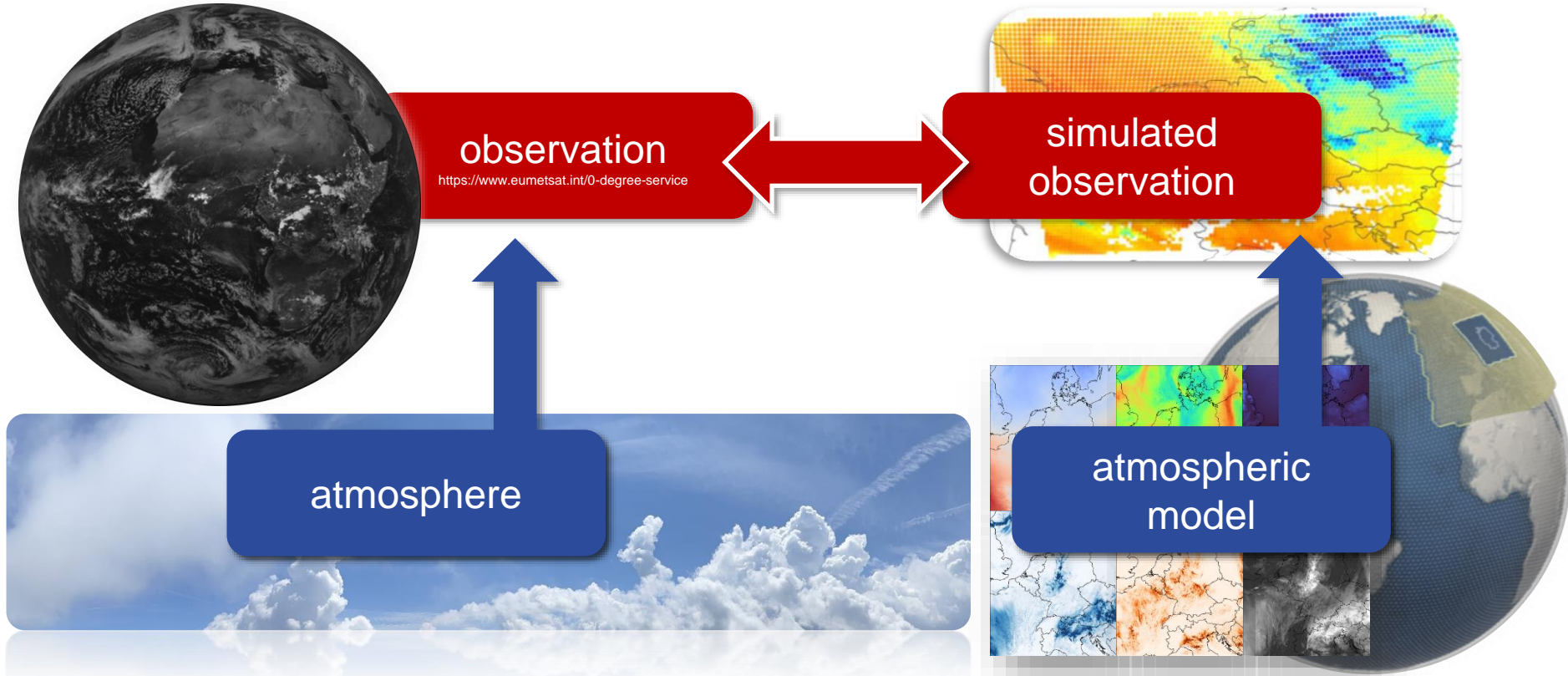


# NWP Model Chain

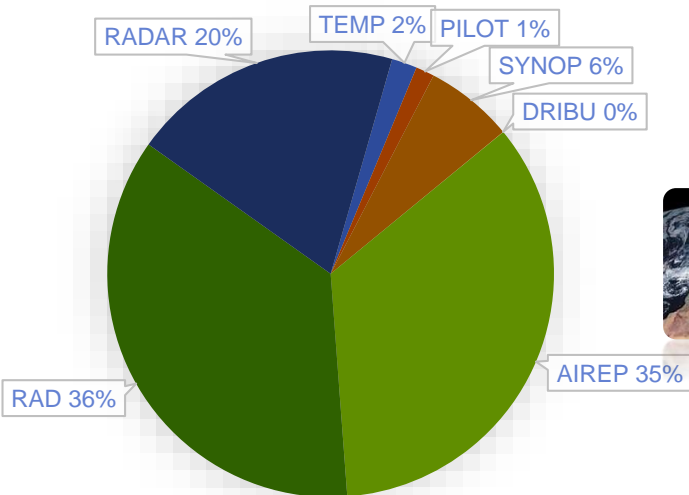
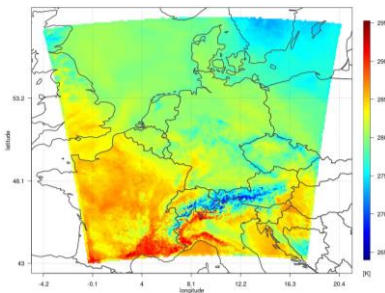
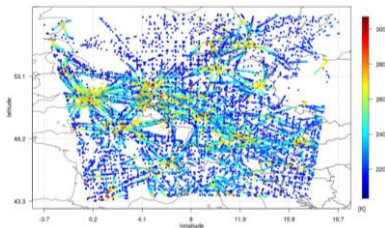




# Observations and NWP Model

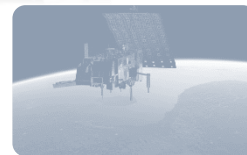


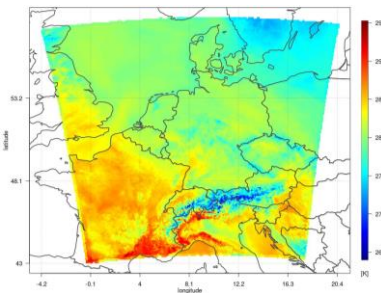
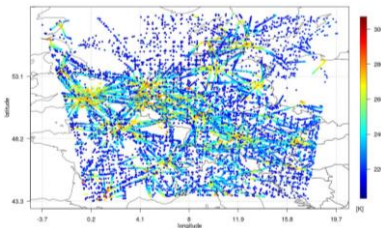
# Operational Data Usage



## ICON-D2

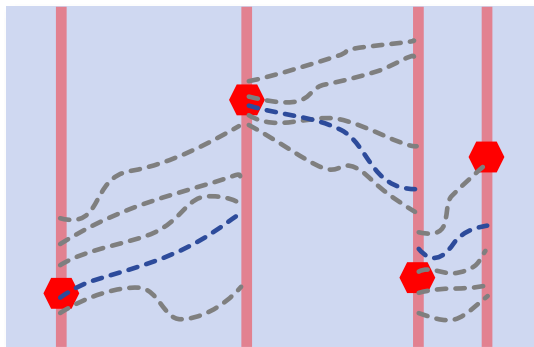
- **operational monitoring:** ~ 10 M data per day (depends on weather)
- **used data:** ~ 2 M data per day





Observations and forecasts have systematic and random errors.

## Localized Ensemble Transform Kalman Filter



- needs observations, observation equivalents, and mean deviation in observation space
- influence of observations is locally restricted
- basic observation operator: interpolation and variable transformation (NetAtmo)
- complex observation operator: EMVORADO (CML data)
- assumptions: observations are unbiased and have Gaussian observation error

Multiple forecasts (ensemble) provide information about the forecast accuracy (spread).



# Observation Processing

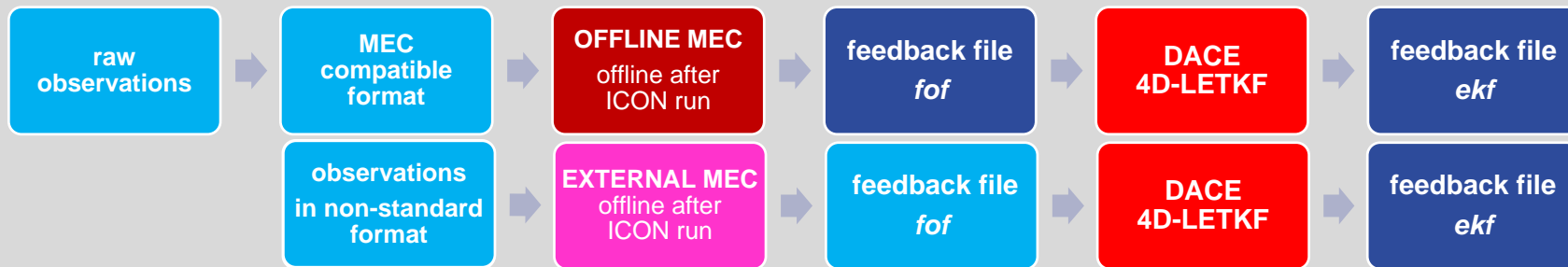
## Online: operational workflow



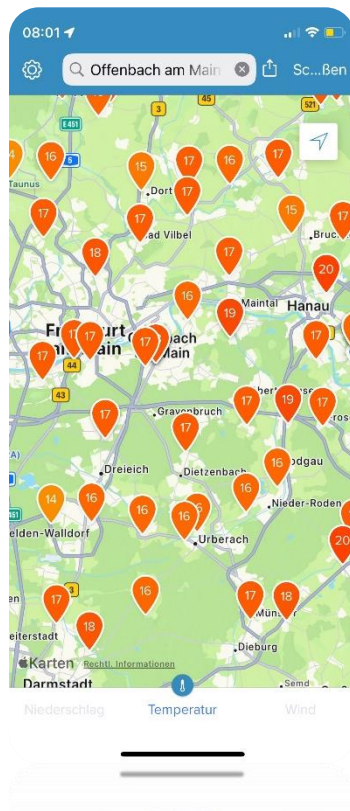
## Online: pseudo operational workflow



## Offline: experimental workflows



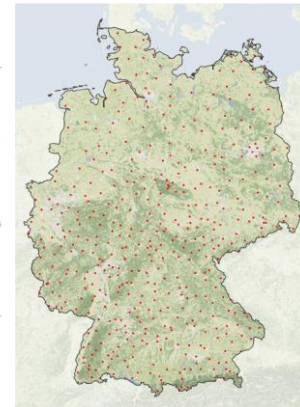
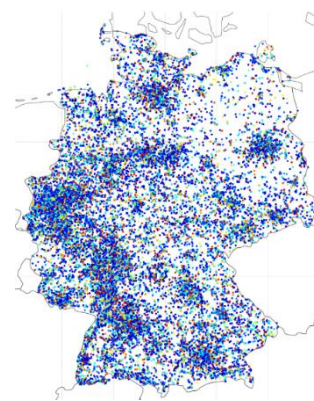
# NetAtmo Data Assimilation



## NetAtmo Project (2020-2021)

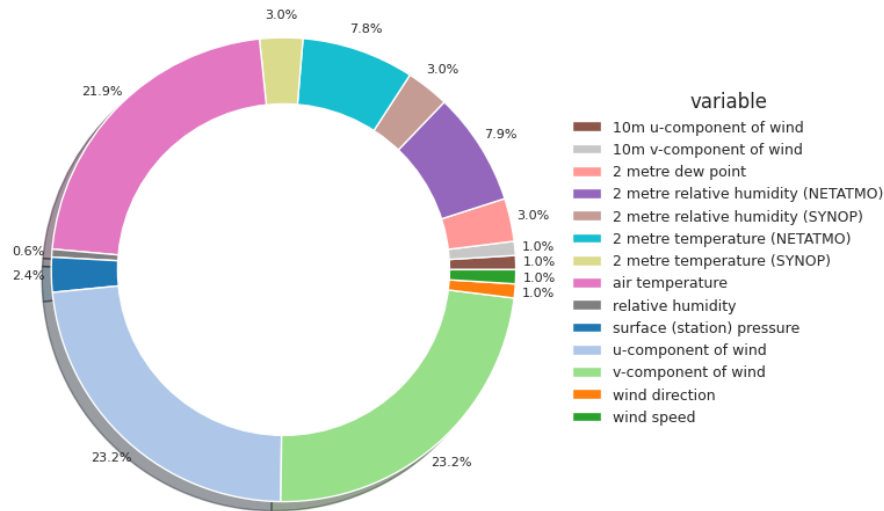
Comparison of the assimilation of SYNOP and NetAtmo stations

Christine Sgoff, Walter Acevedo,  
Zoi Paschalidi, Sven Ulbrich,  
Thomas Kratzsch, Roland Potthast



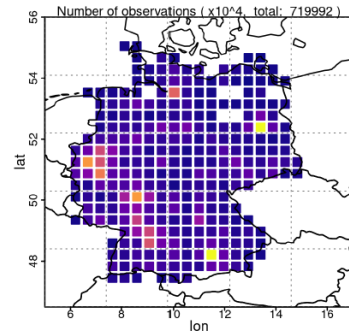
NetAtmo and CML Data Assimilation

# Data coverage

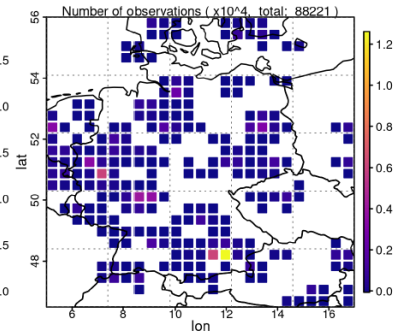


The high density of NetAtmo stations requires spatial thinning.

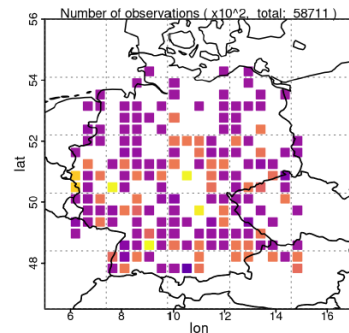
NetAtmo



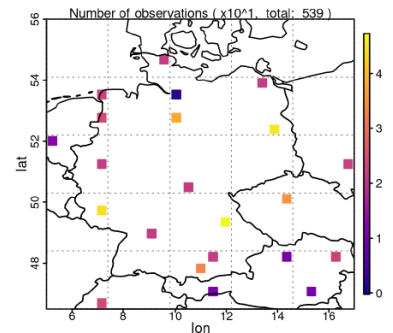
Aircrafts



SYNOP

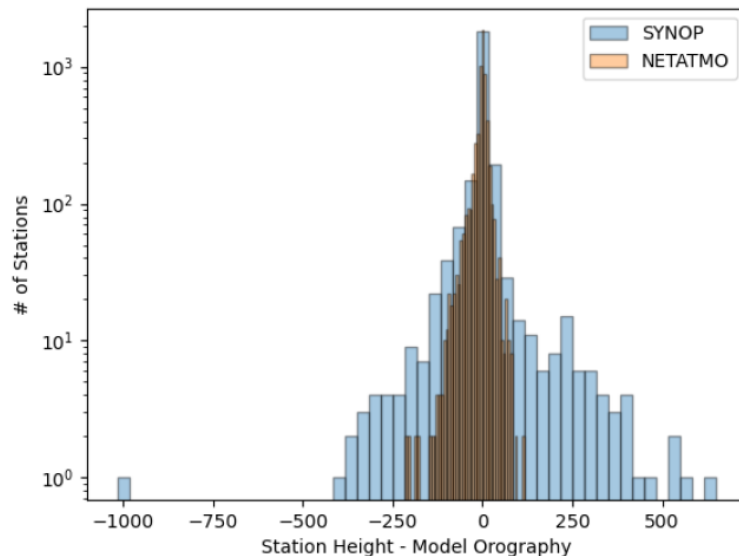
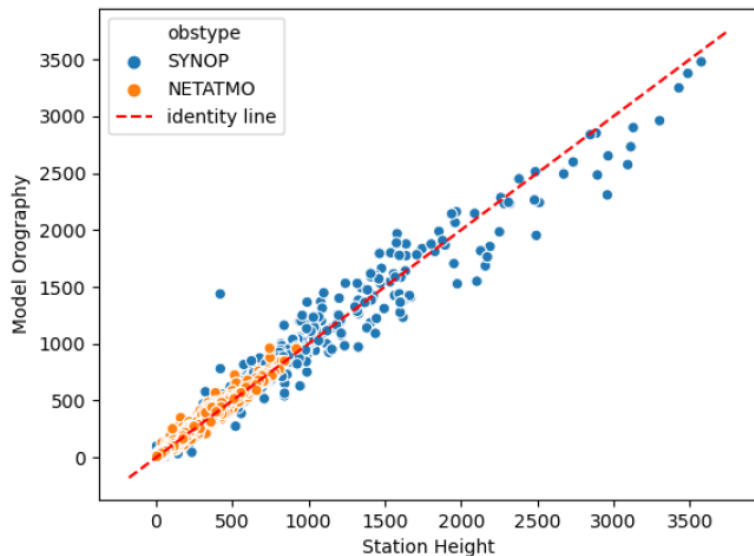


Radiosondes





# NetAtmo – Data coverage



SYNOP stations are evenly distributed across Germany, whereas NetAtmo stations are mainly found in urban metropolitan areas. This poses challenges for both modeling and verification.

Experimental period: 2 weeks in September 2018

## Quality control:

### preprocessing control (observations only)

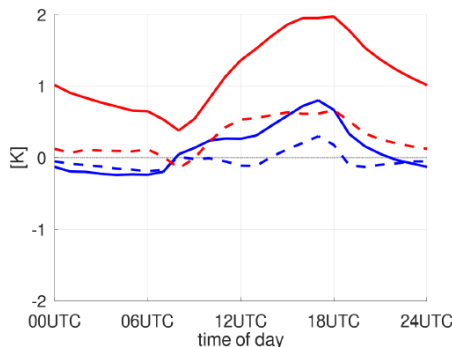
- plausibility control (feasible variable range)
- double records control

### processing control (with ICON model data)

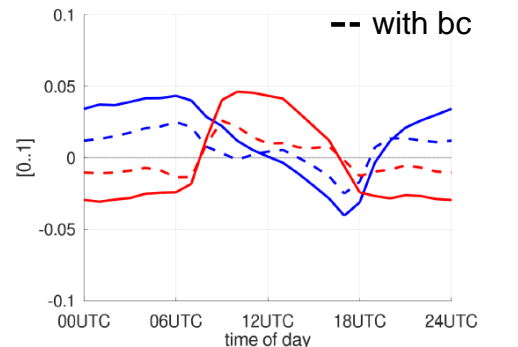
- first guess control
- modelling check (height control)
- error checks

## Bias correction

### 2m humidity



### 2m temperature



NetAtmo data have a significantly larger bias than SYNOP data.

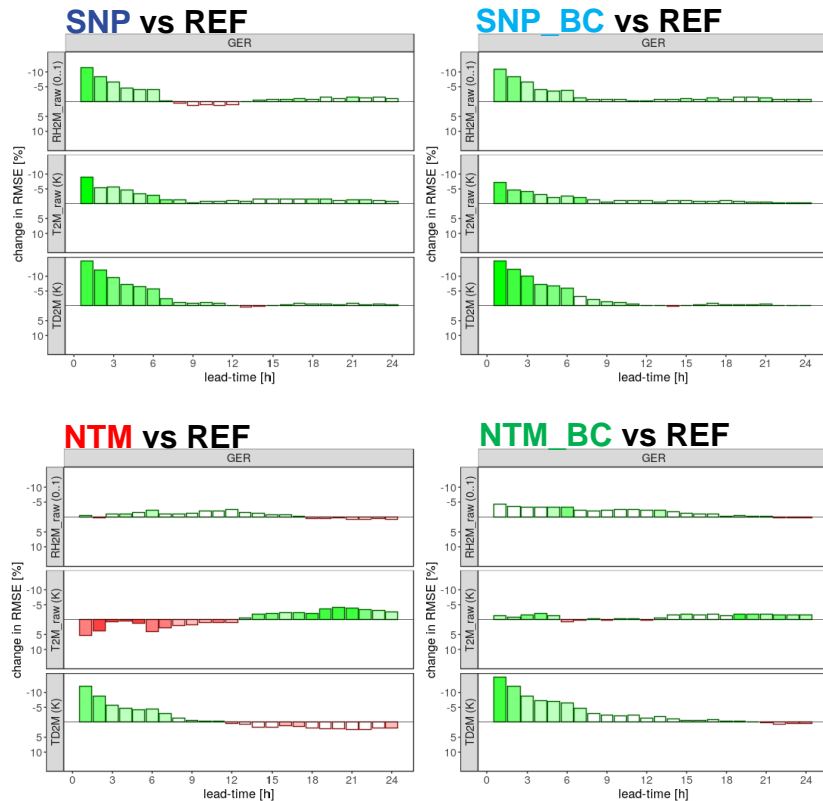
An appropriate bias correction attenuates the diurnal cycle.  
input for NetAtmo: day time  
input for SYNOP: day time and cloud cover

# NetAtmo – Assimilation experiments

experiment name	assimilated observations	bias correction
REF	Conventional (the operational set up)	no
SNP	Conventional + synoptic T2M+RH2M	no
SNP_BC	Conventional + synoptic T2M+RH2M	yes
NTM	Conventional + NetAtmo T2M+RH2M	no
NTM_BC	Conventional + NetAtmo T2M+RH2M	yes



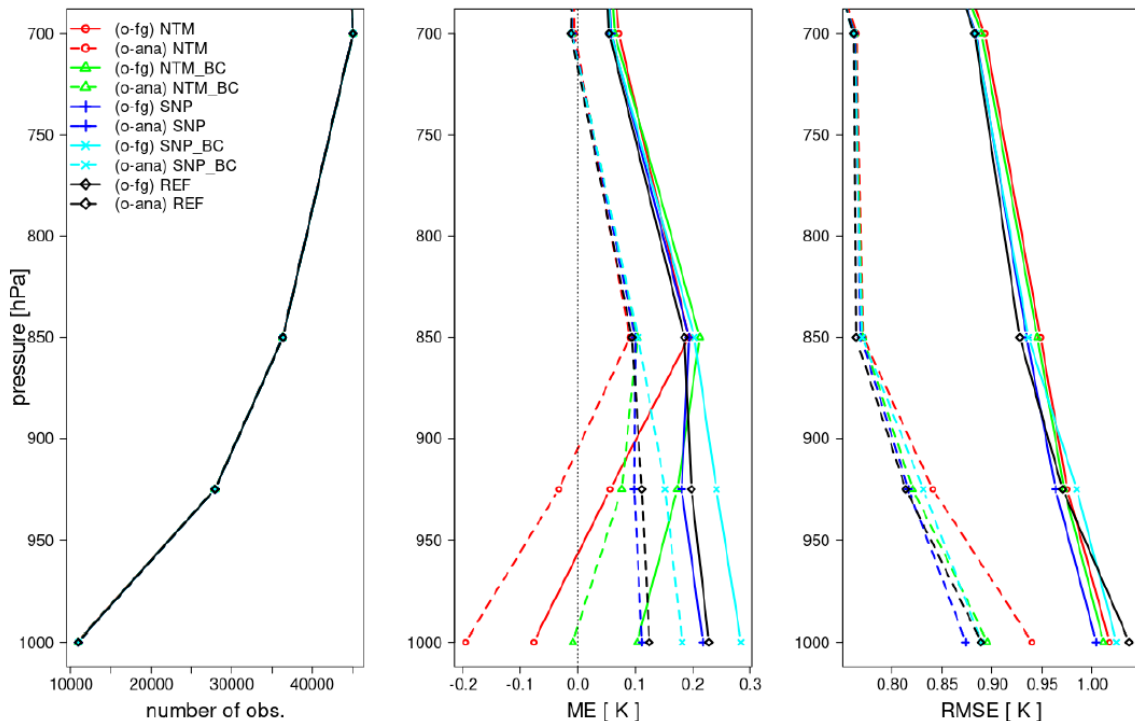
# NetAtmo – SYNOP verification



name	observations	bias correction
REF	Conventional	no
SNP	Conventional + SYNOP	no
SNP_BC	Conventional + SYNOP	yes
NTM	Conventional + NetAtmo	no
NTM_BC	Conventional + NetAtmo	yes

# NetAtmo – Upper air verification

## Upper air temperature: Analysis and first guess

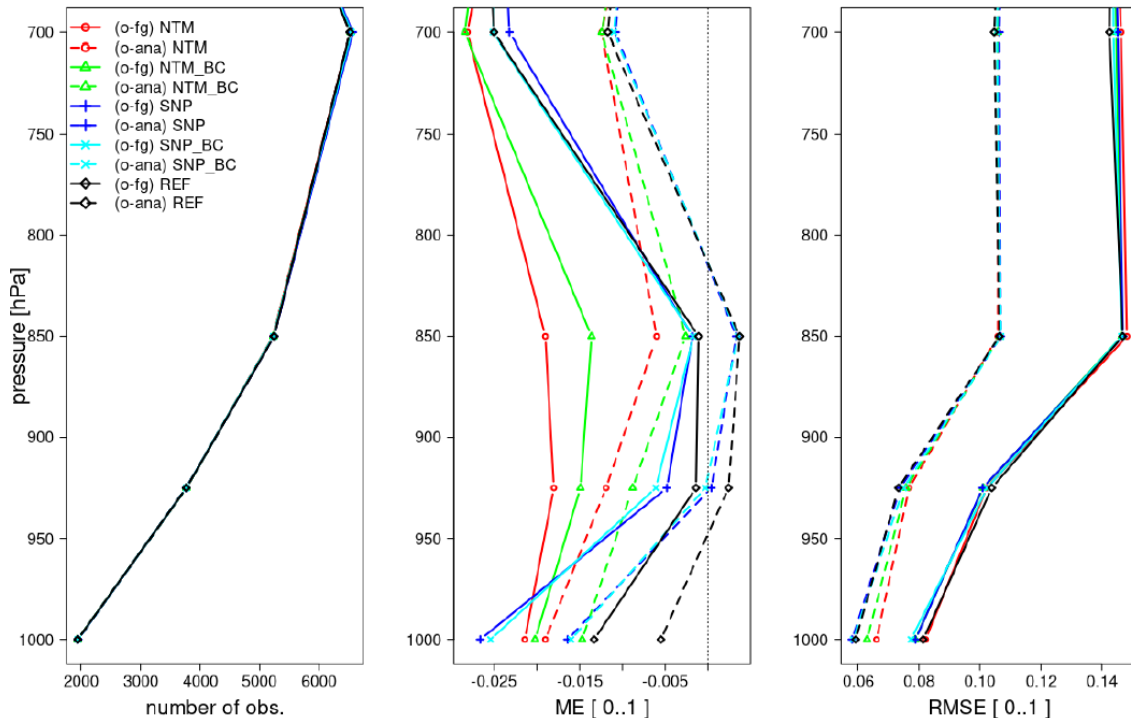


name	observations	bias correction
REF	Conventional	no
SNP	Conventional + SYNOP	no
SNP_BC	Conventional + SYNOP	yes
NTM	Conventional + NetAtmo	no
NTM_BC	Conventional + NetAtmo	yes

— first guess  
-- analysis

# NetAtmo – Upper air verification

## Upper air humidity: Analysis and first guess



name	observations	bias correction
REF	Conventional	no
SNP	Conventional + SYNOP	no
SNP_BC	Conventional + SYNOP	yes
NTM	Conventional + NetAtmo	no
NTM_BC	Conventional + NetAtmo	yes

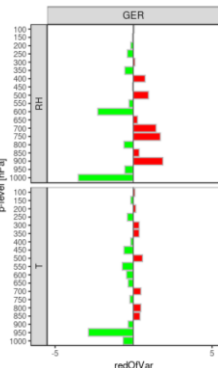
— first guess  
-- analysis



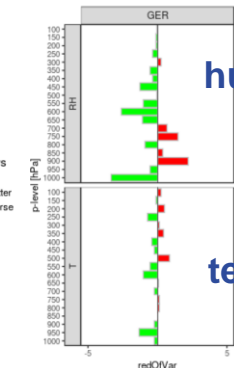


# NetAtmo – Upper air verification

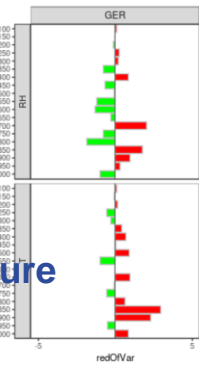
SNP vs REF



SNP\_BC vs REF



NTM vs REF

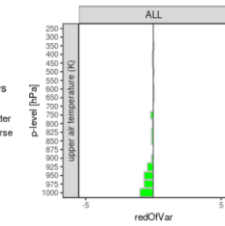
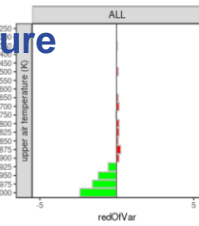
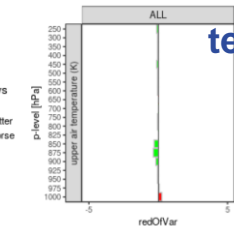
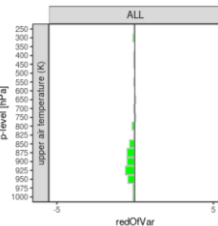


NTM\_BC vs REF



humidity

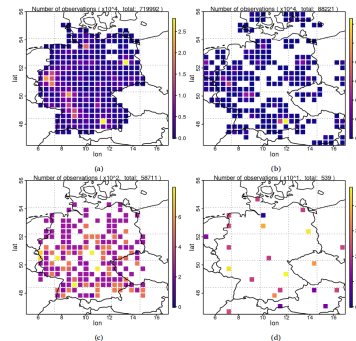
temperature



temperature

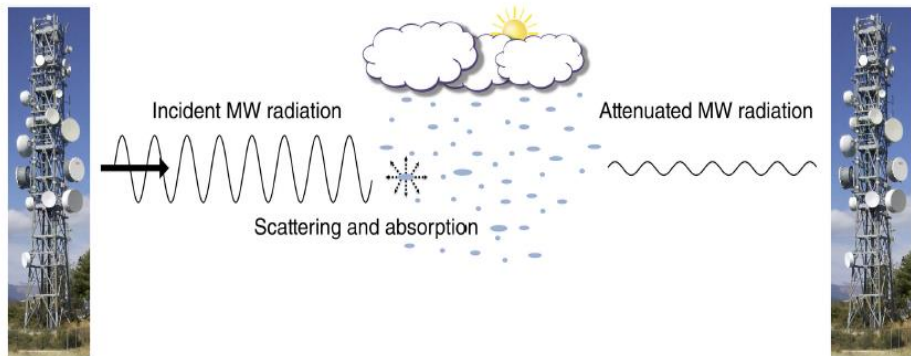
name	observations	bias correction
REF	Conventional	no
SNP	Conventional + SYNOP	no
SNP_BC	Conventional + SYNOP	yes
NTM	Conventional + NetAtmo	no
NTM_BC	Conventional + NetAtmo	yes

verification of 24h forecasts



NetAtmo and CML Data Assimilation





<https://doi.org/10.1002/wat2.1337>

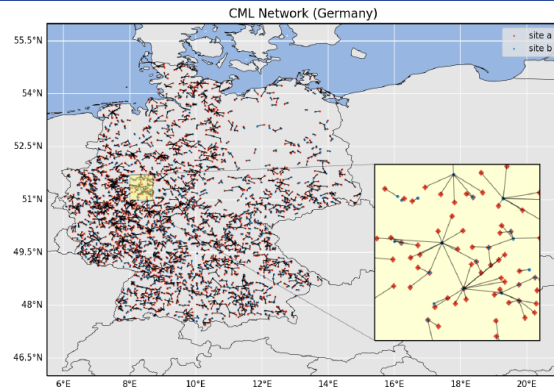
## Commercial Microwave Links (CMLs)

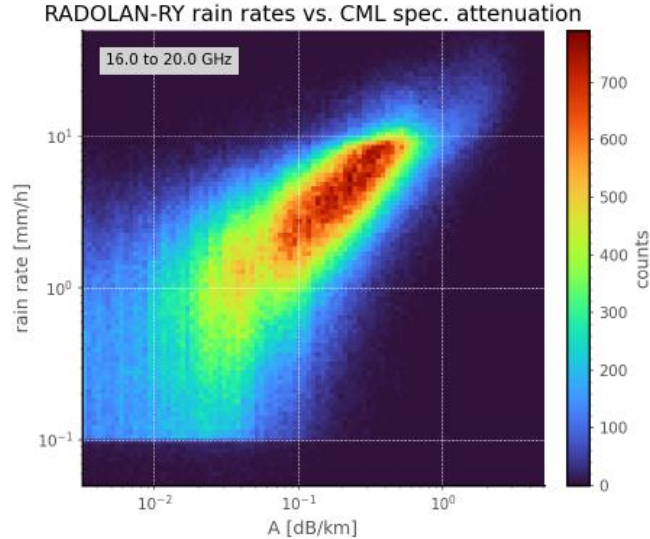
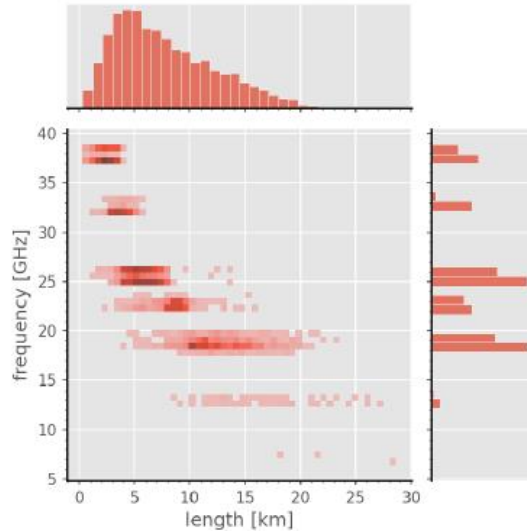
- wireless communication connections between mobile phone towers that transmit data using microwave frequencies
- sensitive to atmospheric conditions, especially **rain**
- dataset of ~4000 CMLs (June 2019)  
one minute resolution

## RealPEP (2021-2025)

Commercial Microwave Link Data Assimilation with the LETKF

Klaus Vobig, Roland Potthast,  
Christian Chwala, Julius Polz

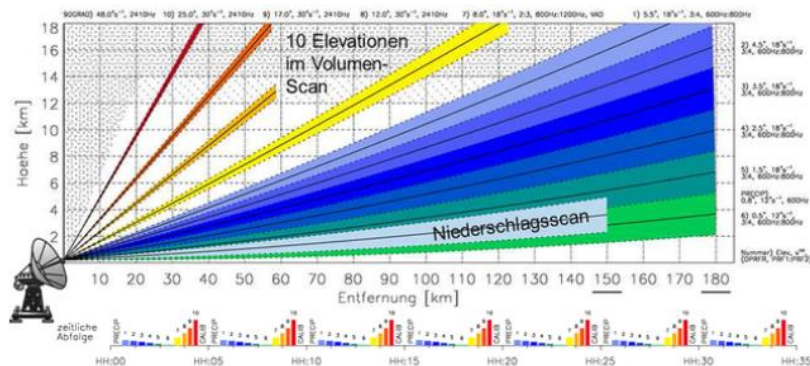
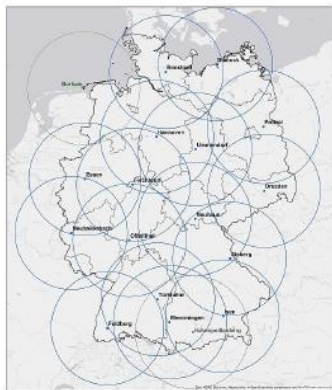




## Commercial Microwave Links (CMLs)

- CML frequency between 10 and 40 GHz
- power law relation between rainfall rate and CML attenuation
- assimilation of path-averaged specific attenuations

# Radar Operator EMVORADO

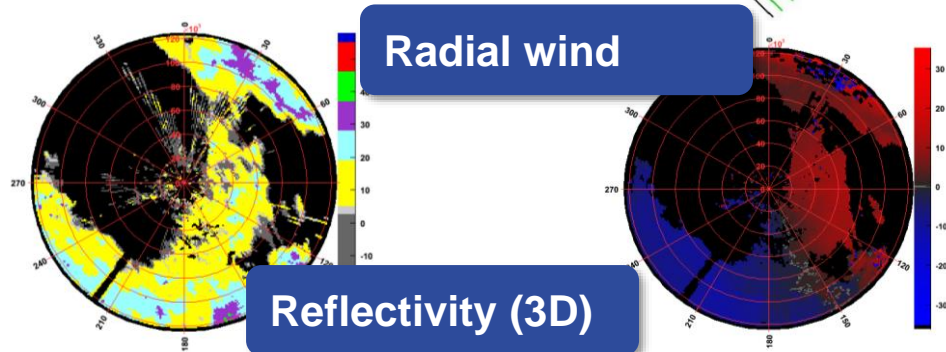


## Efficient Modular Volume scan Radar Operator

- framework for simulation of radar volume data of entire networks in a highly configurable way
- computes reflectivities and attenuations

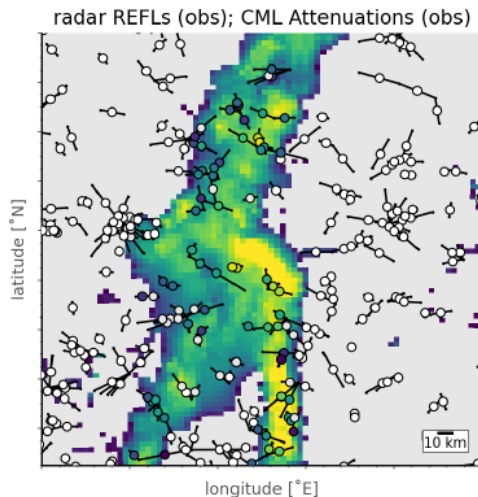
## German Radar Network

- 17 dual-polarization C-band Doppler Radars
- reflectivity and radial wind assimilated in ICON\_KENDA, and Latent Heat Nudging is employed

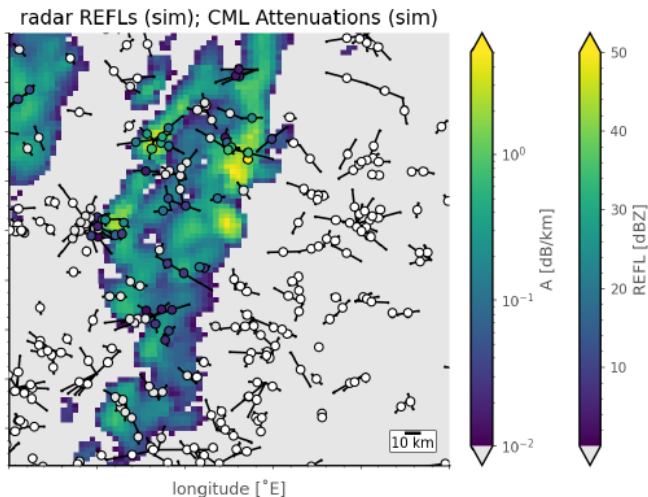




## observations



## simulations



## Radar

- 17 stations, many azimuths for certain elevations
- fixed frequency (C-band, ~5 GHz)
- standard EMVORADO setting

## CML

- 4000 stations, fixed individual azimuth and elevation
- individual frequencies (10-40 GHz)
- special EMVORADO setting to compute the path-integrated one-way attenuation
- experimental data assimilation workflow

Single case studies

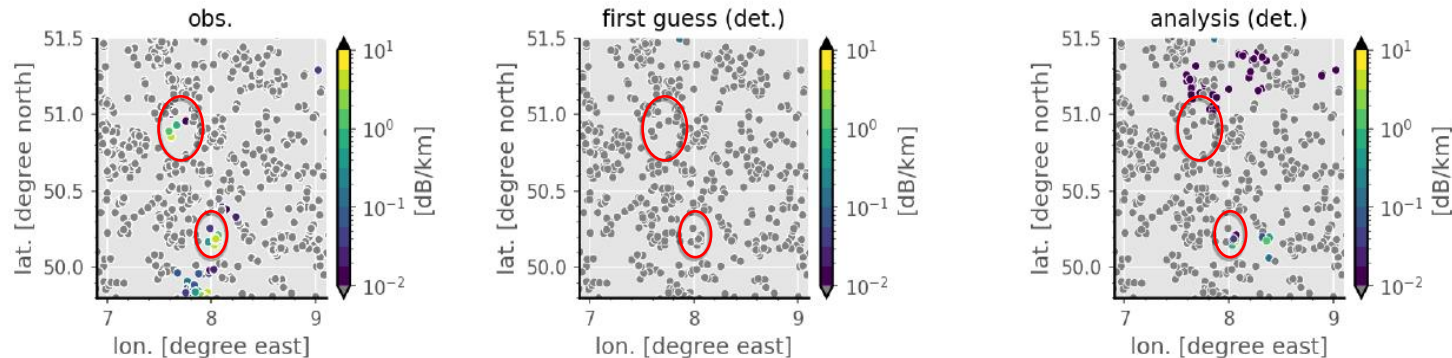
## **preprocessing control (observations only)**

- temporal average over 10 min before assimilation time
- spatial assignment to the center of the link path

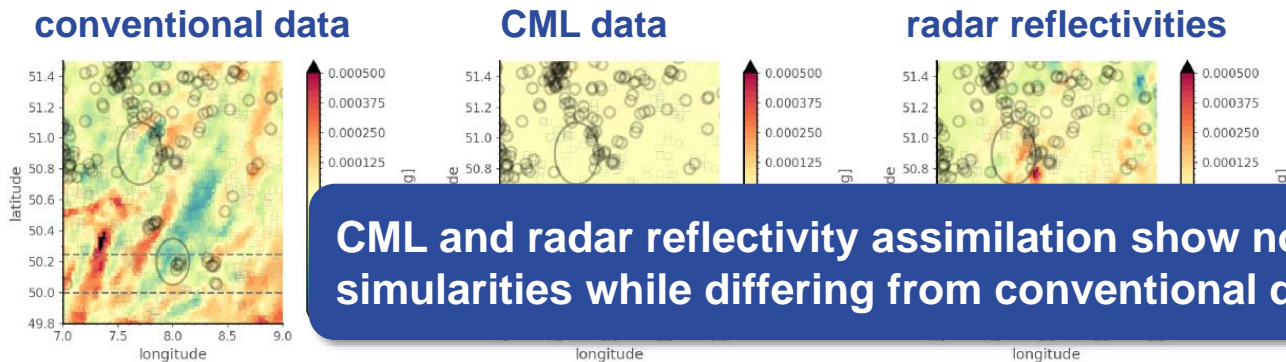
## **processing control (with ICON model data)**

- range control (trim small values in observations and simulations)
- ensemble checks
- dynamic observation error (1 dB / path length)

# CML – Single case study

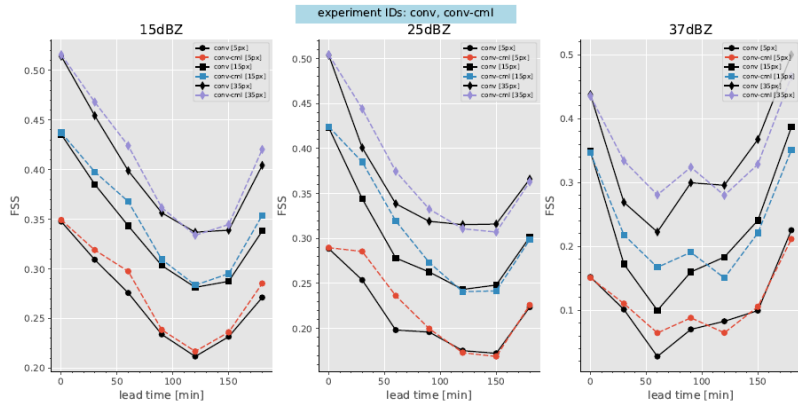


humidity increments (vertically averaged) from different observing systems

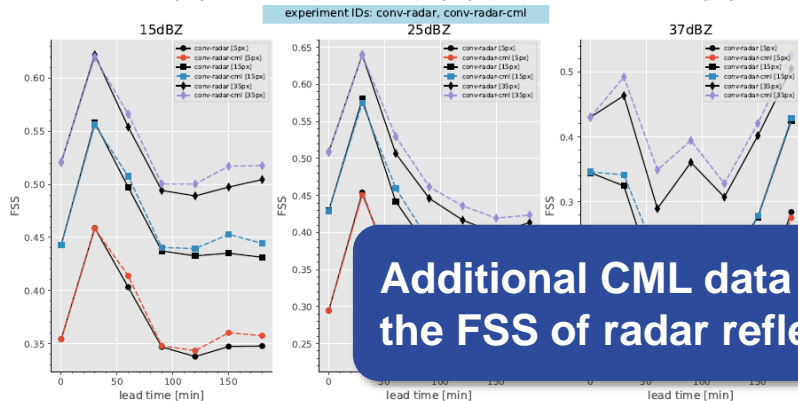


**CML and radar reflectivity assimilation show notable similarities while differing from conventional data assimilation.**

# CML – Forecast impact



**Conventional data only (black)**  
compared to  
conventional and CML data  
for different scales (different colors)



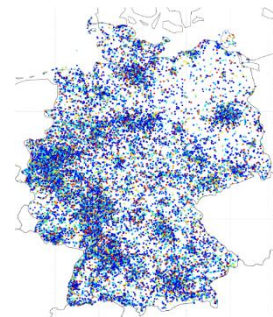
**Conventional and radar data only (black)**  
compared to  
conventional, radar, and CML data  
for different scales (different colors)

**Additional CML data assimilation is capable of improving the FSS of radar reflectivity forecasts.**

**Crowdsourcing and opportunistic observations have the potential to enhance numerical weather prediction (NWP) within regional forecasting systems.**

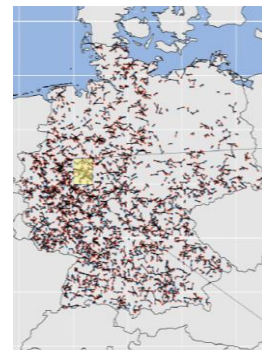
## NetAtmo data

- Bias correction is essential.
- While the quantity of these observations may compensate for their lower quality compared to SYNOP data, this requires sophisticated treatment in the data assimilation (horizontal superobbing, horizontal observation error correlations)
- If implemented properly, NetAtmo data can provide a valuable densification of the observational network.



## CML data

- CML data integration would require substantial and fundamental developments in observation modeling.
- CML data are likely redundant when compared to radar observations; however, CML data may be an adequate substitute where no radar data are available.





# Enhancing Numerical Weather Prediction with Opportunistic and Crowdsourced Observations: Insights from NetAtmo and Microwave Link Assimilation

Stefanie Hollborn  
FE12 – Observation Modelling and Verification  
FE1 – Meteorological Analysis and Modelling



- Christine Sgoff, Walter Acevedo, Zoi Paschalidi, Sven Ulbrich, Elisabeth Bauernschubert, Thomas Kratzsch, Roland Potthast: Assimilation of crowd-sourced surface observations over Germany into a Regional Weather Prediction System, QJRM, 2022, <https://doi.org/10.1002/qj.4276>
- RealPEP project: <https://www2.meteo.uni-bonn.de/realpep/doku.php?id=home>
- Klaus Vobig, Roland Potthast, Christian Chwala, Julius Polz: Commercial Microwave Link (CML) Data Assimilation with the LETKF, in preparation.

