

KENDA : Km-Scale Ensemble-Based Data Assimilation *(Schraff et al. 2016, QJRMMS)*

currently based on 4-D LETKF + latent heat nudging

- some recent KENDA developments

- ICON-LAM (Limited Area Mode) vs. COSMO , 3-D vs. 4-D LETKF

(Hendrik Reich, Christian Welzbacher, Christoph Schraff, Harald Anlauf, Klaus Stephan, Roland Potthast et al., DWD)

- 3-D radar reflectivity in LETKF *(Thomas Gastaldo, Virginia Poli, ARPAE Bologna)*

- GPS Slant Total Delay *(Michael Bender et al., DWD)*

- crowd-sourced data

- project on car data, Ultra-Rapid DA (URDA)

(Walter Acevedo, Zoi Paschalidi et al., DWD)

COSMO

$H(x) \rightarrow H(x(t_{obs}))$
during model integration
currently 1 hour

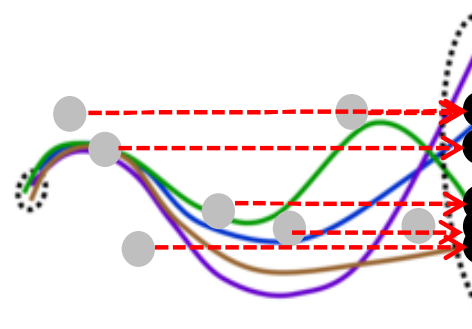
4D



→ full 4-D (o-b, P_b) (online)

$H(x) \rightarrow H(x_{FG}(t_a))$

3D

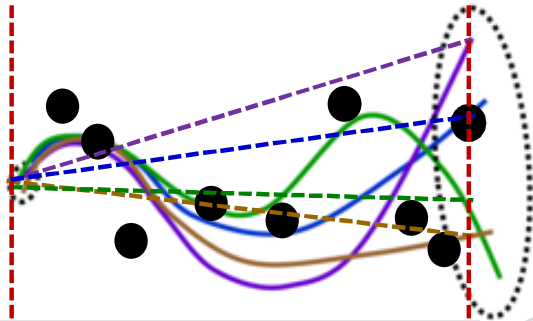


→ 3-D (o-b, P_b) (MEC)

ICON-LAM vs. COSMO (3-D this talk)

$H(x) : x(t_{obs})$ by interpolation betw. $x_{ana}(t_a-1h)$ and $x_{FG}(t_a)$

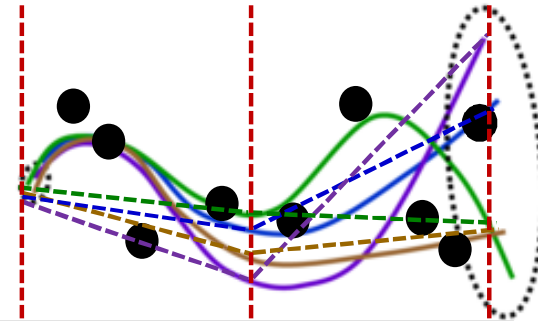
4D



→ interpol. 4-D (2 t.lv.)

$H(x) : x(t_{obs})$ by interpolation betw. $x_{ana}(t_a-1h)$, $x(t_a-1/2h)$, $x_{FG}(t_a)$

4D



→ interpol. 4-D (3 t.lv.)

ICON-LAM (x time lev.)

ICON-DE vs. COSMO-DE with LHN

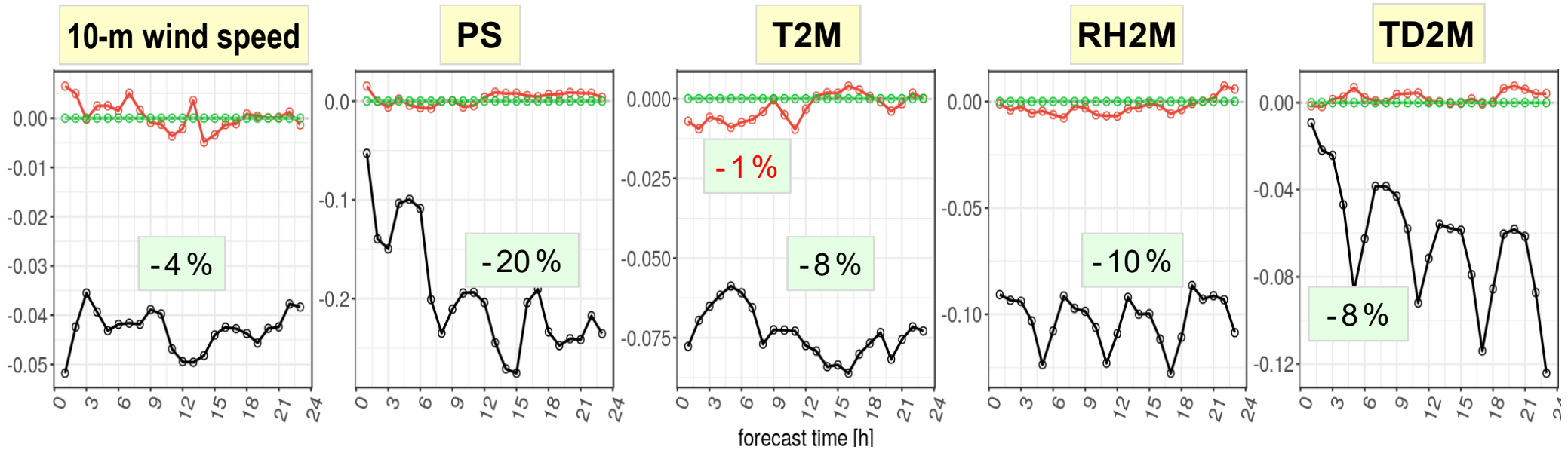
(27 May – 10 June 2016)



Deutscher Wetterdienst



SYNOP verification
relative change of RMSE
vs. COSMO-DE 3-D

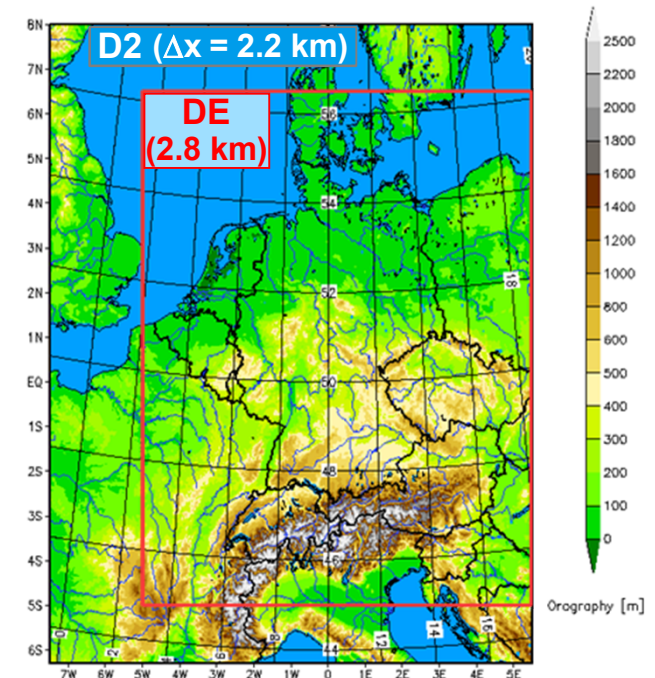


ICON-DE 3-D
COSMO-DE 3-D
COSMO-DE 4-D

similar error reductions
in terms of std. dev.
(except for PS where
both std and bias are
strongly reduced)

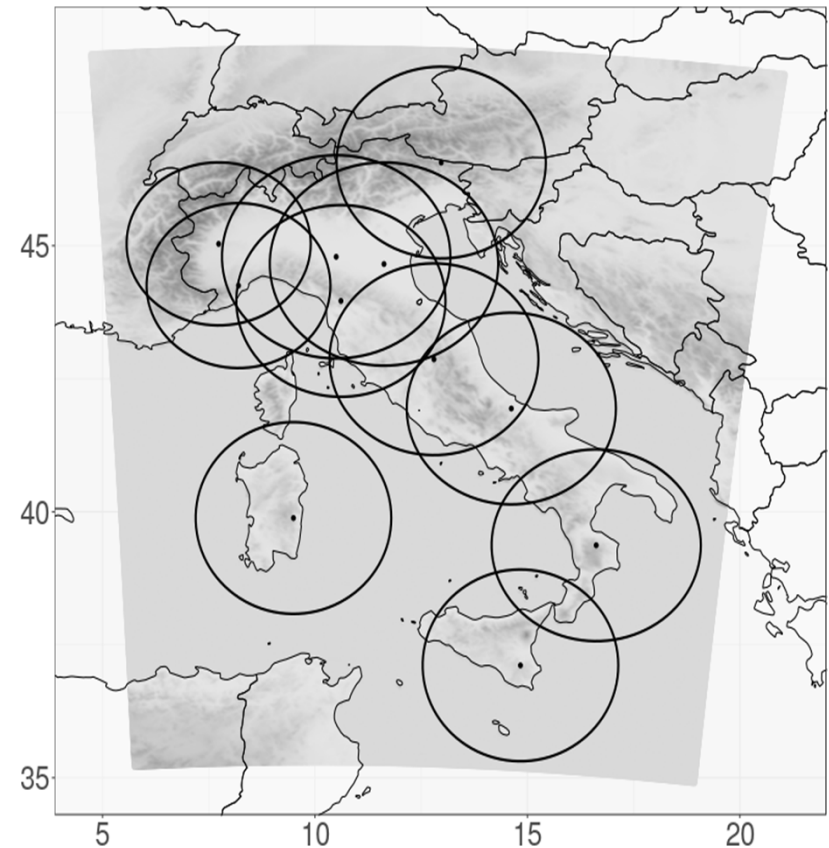


- **3-D** LETKF *only slightly* worse than **4-D** LETKF (1-h LETKF cycling)
- **interpolated 4-D** LETKF (\rightarrow ICON-LAM) \sim as good as online 4-D LETKF
(2-week convective summer period)
- (**3-D** LETKF + LHN:) **ICON-DE** much better than COSMO-DE in most variables
(upper-air troposphere; wind, T also better, RH slightly worse ;
precip \sim neutral (LHN needs further tuning))
(summer & winter periods)
- **4-D** LETKF ICON-D2 ($\Delta x = 2.2$ km)
experiments underway
- Plan: **ICON-D2**
 - pre-operational parallel suite autumn 2019
 - operational 12 / 2020



Setup

- 1-hrly LETKF 2.2 km
- reference: conv. obs (TEMP, SYNOP, aircraft, wprof) in LETKF
+ continuous **LHN**
- **Radar Z:** superobbing 10 km,
1 scan per hour (closest to analysis time,
i.e. 3-D approach for radar obs)
(replacing LHN)
- evaluated: deterministic forecast every 3 hrs
3 periods (72 – 123 forecasts)

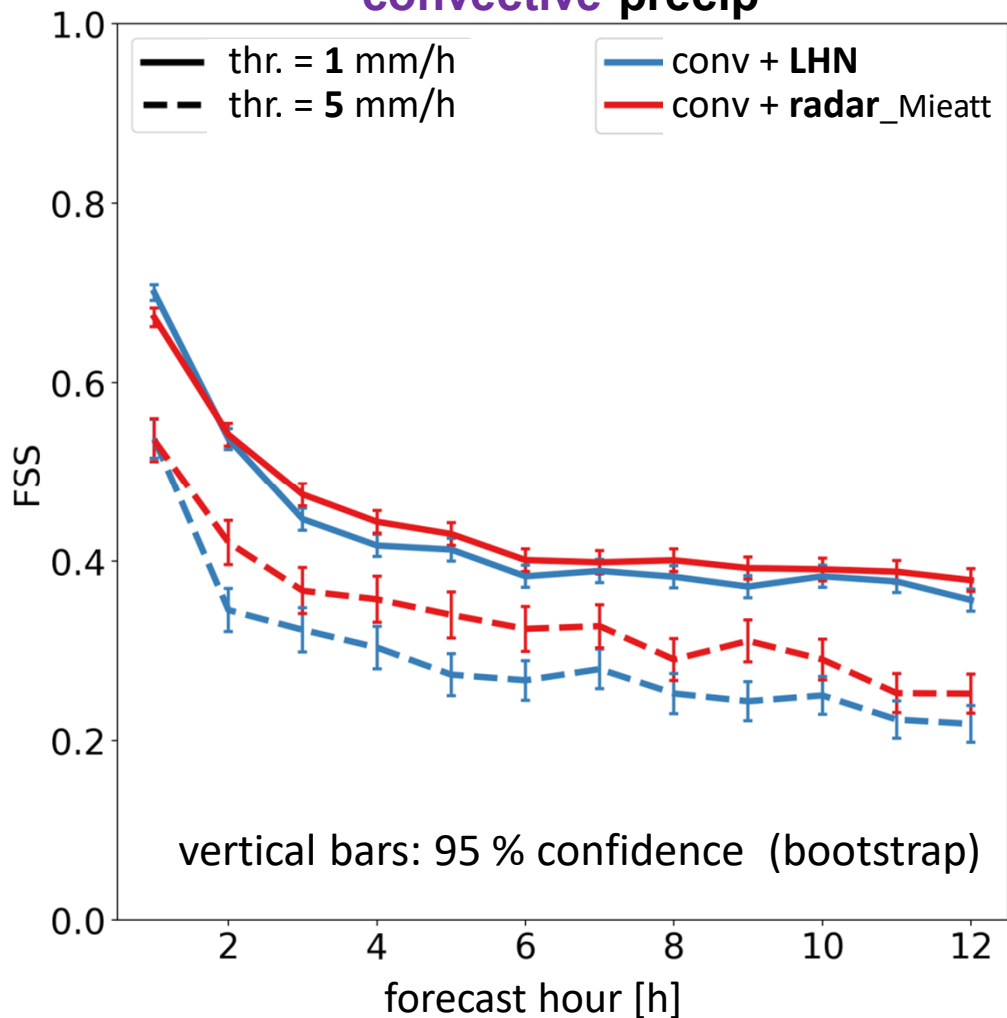


Event	Start	End	Type of event
Sept. 2018	31/08 - 00UTC	09/09 - 00UTC	thunderstorms
Oct. 2018	30/09 - 15UTC	14/10 - 00UTC	organized thunderstorms
Nov. 2018	26/10 - 12UTC	11/11 - 00UTC	stratiform precipitation

FSS (0.2°) for 1-h precip vs. radar composite adjusted by rain-gauges

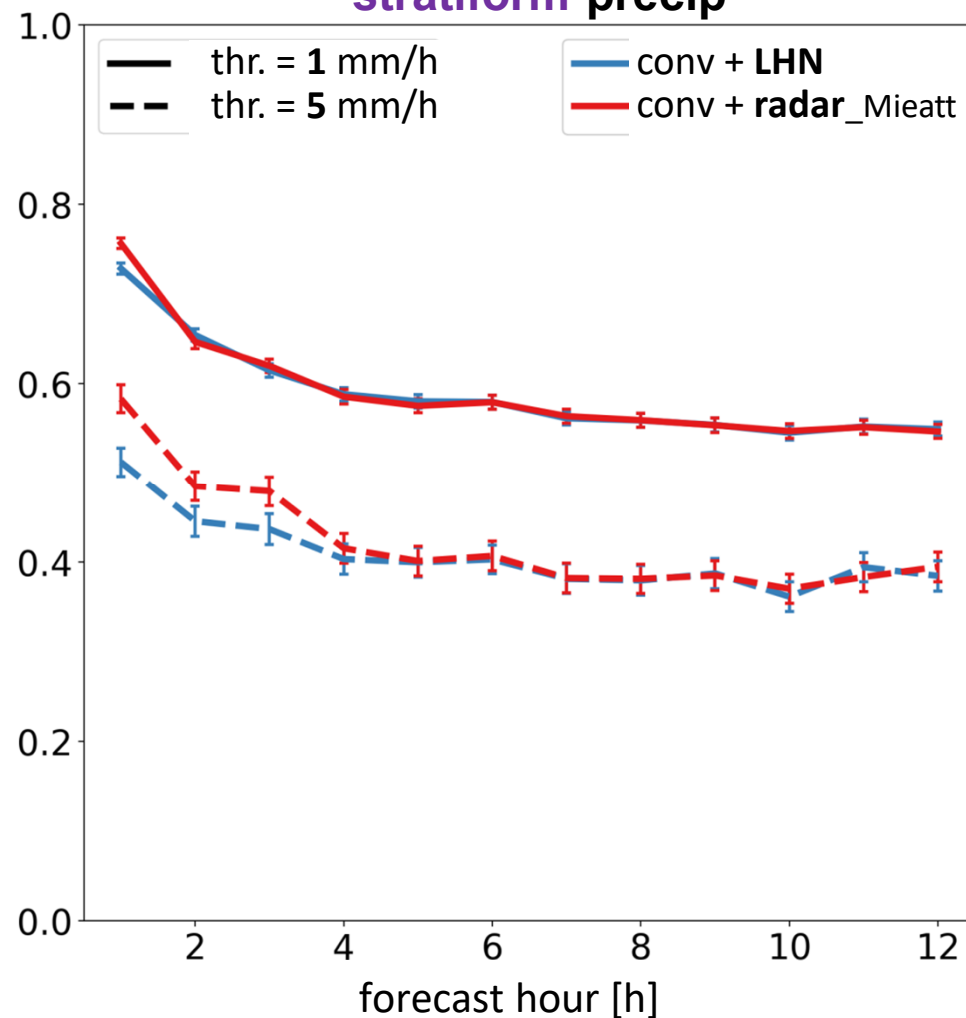
1.5 weeks Sept. + 2 weeks Oct. 2018

convective precip



2 weeks November 2018

stratiform precip



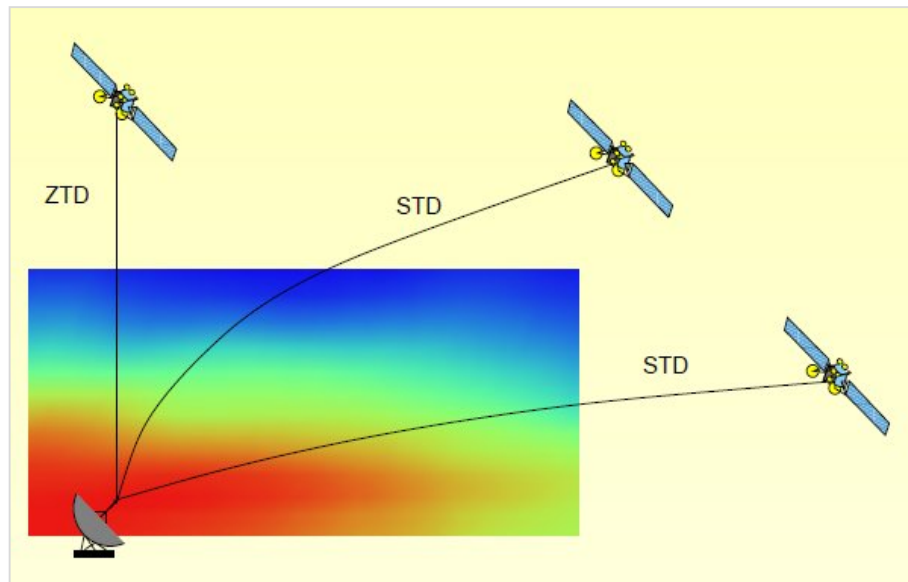
Results at ARPAE: assimilation 3-D reflectivity by LETKF compared to LHN

- **precip:** statistically **significantly better**, especially for heavy and non-organized precip;
upper-air: better RMSE (T, RH, mainly **wind**),
surface: similar (T2M slightly improved, RH2M + ps slightly degraded, 10-m wind neutral)
- plan: replace LHN by 3-D radar reflectivities in operational suite next year

DWD:

- implemented in ICON-LAM, tests underway (will not be made operational in COSMO)

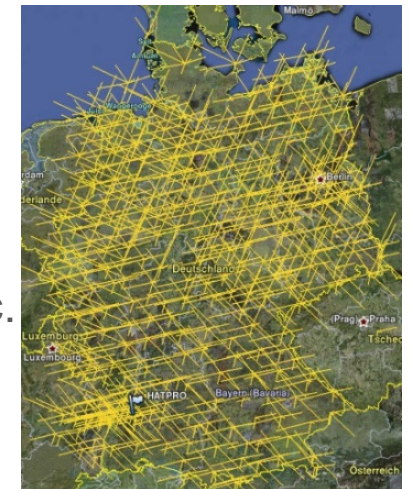
GPS (GNSS) STD: **humidity integrated over (slant) path**
from ground station to GNSS (GPS) satellite, all weather obs



- STD (slant total delay), elevation $\varepsilon = 7^\circ - 90^\circ$
→ after thinning ~ 2000 obs/h
- ZTD (zenith total delay):
mapping of delays to 1 zenith info,
 < 300 obs/h

- 300 stations in COSMO-DE domain, thinning (3-D approach)
- no consistent positive impact → improved bias correction, bug fixes, etc.
- many higher-elevation STD provide info similar to each other & to ZTD
(→ redundant info)

→ new experiment: assimilate **ZTD + low-elevation STD** ($7^\circ < \varepsilon < 25^\circ$)



GPS STD convective summer period

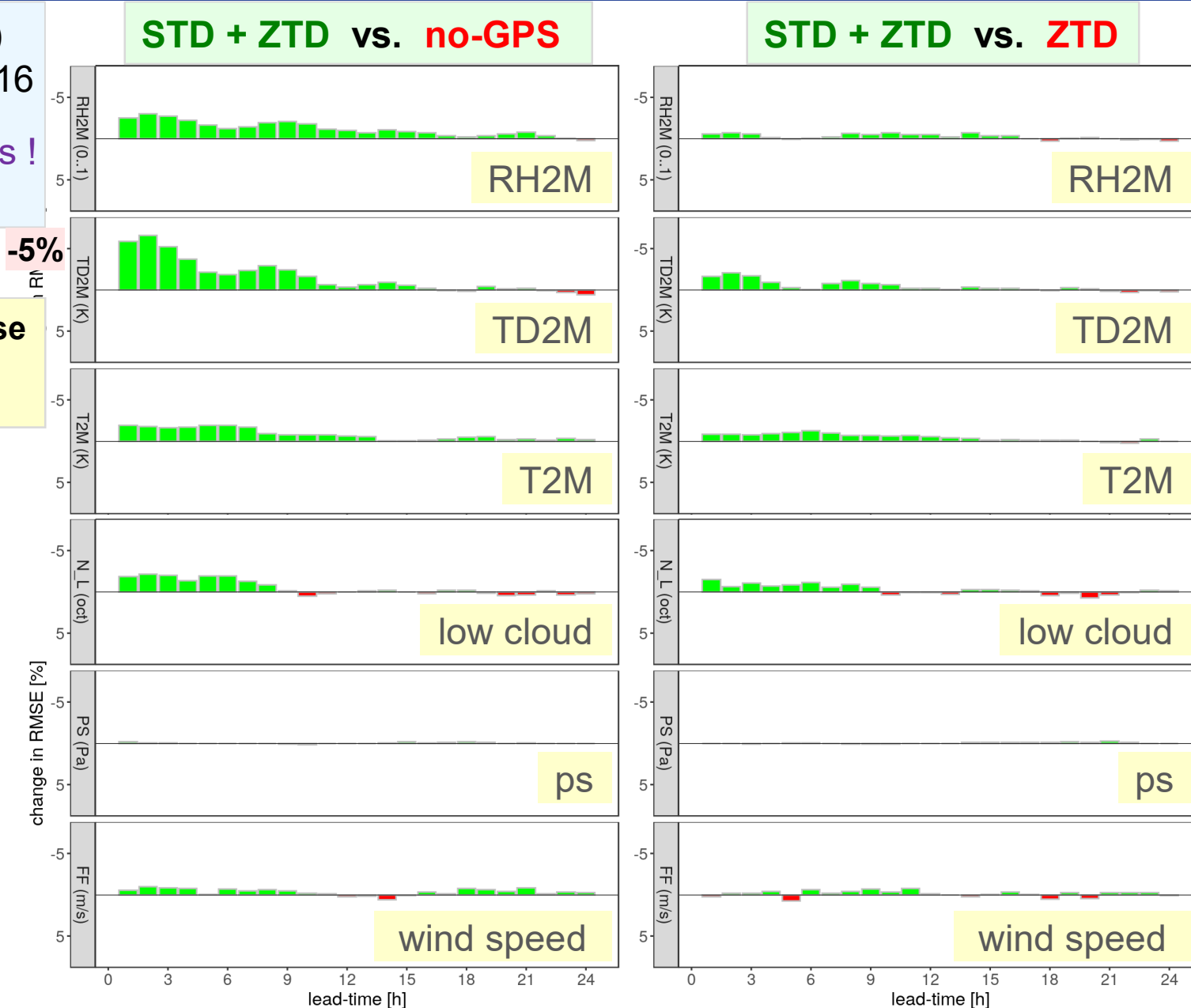
Synop verif. (GER)
25 May – 1 July 2016



→ 5 weeks !

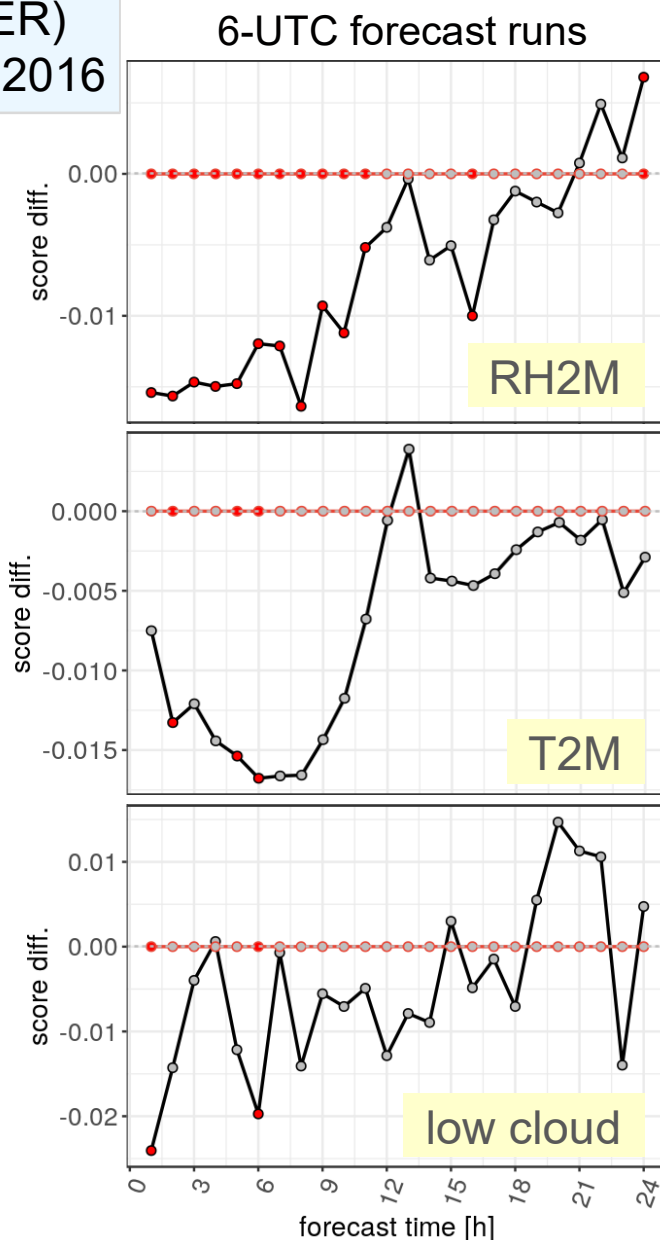
% change of rmse
(as function of
lead time)

 better
 worse



GPS STD convective summer period

Synop verif. (GER)
25 May – 1 July 2016



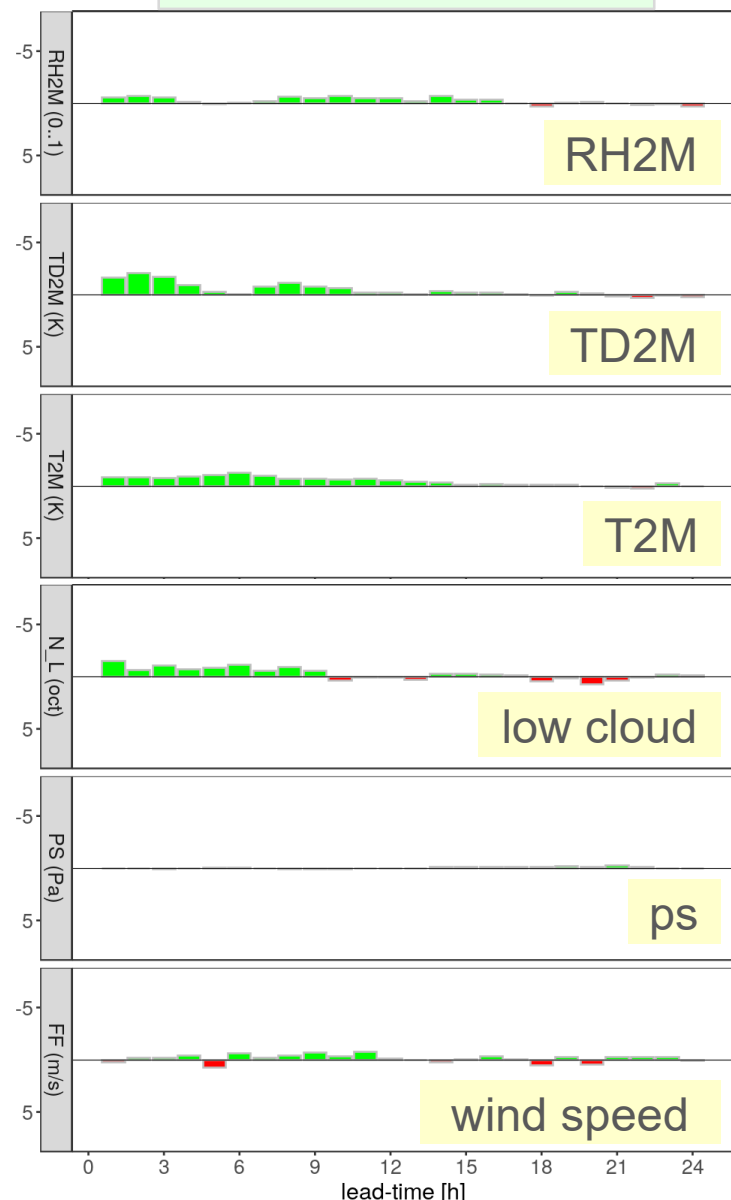
Experiment
 ● 2100.12
 ○ 2100.14

Valid hour
 — ALL

Sig. diff. (95%)
 ○ n.a.
 ○ no
 ● yes

improvements
by adding STD
to ZTD
are
occasionally
significant

STD + ZTD vs. ZTD

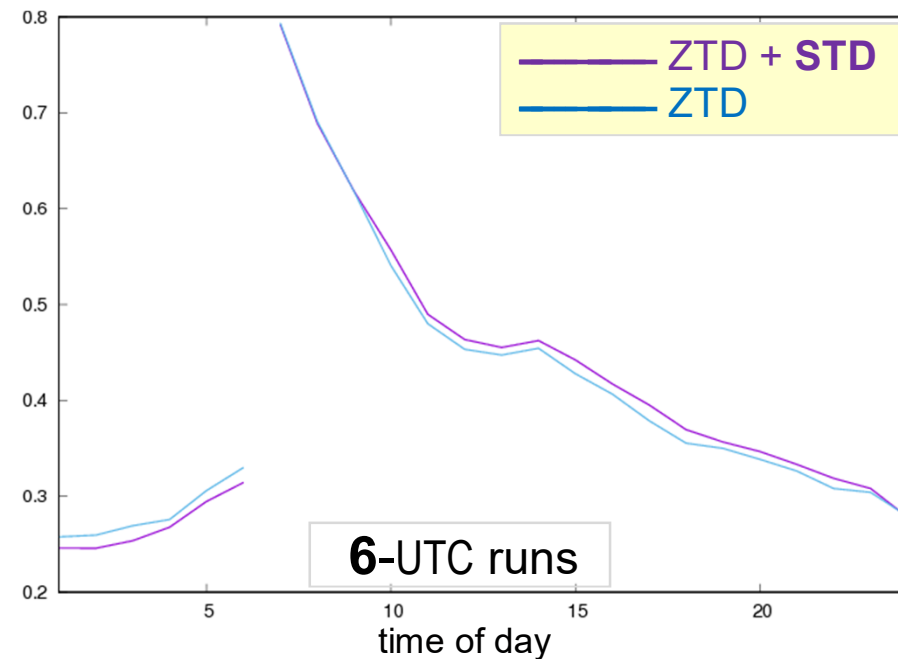


precip vs. radar
26 May – 28 June 2016



→ 5 weeks !

1-hrly precip
FSS
(30 km)
1 mm/h



- moderate positive impact from **ZTD** (upper-air, surface, precip (0-, 6- UTC runs))
- **small extra benefit by** adding (low-elevation) **STD** on top of ZTD, 'occasionally' significant (Synop verif.)
- future: – experiments with ICON-LAM (smaller model biases)
– potential for enhanced benefit by improved **bias correction of STD** ?

- **GPS STD** small extra benefit on top of ZTD
- radar **reflectivity** Z ARPAE: better than LHN, **soon operational** (with COSMO);
DWD: → ICON-LAM
- radar **radial wind** small consistent positive impact after increasing obs errors
where Z low
→ in **parallel suite** for COSMO soon
- **all-sky SEVIRI** IR + VIS work ongoing → ICON-LAM
- **T2M**, RH2M: (at last) resources since spring, work on
conditional non-linear bias correction (*Otkin + Potthast 2019, MWR*)
→ ICON-LAM
- ongoing work on other obs, e.g. wind lidar, Raman lidar, MW radiometer, ...
lightning, land surface temperature, ...

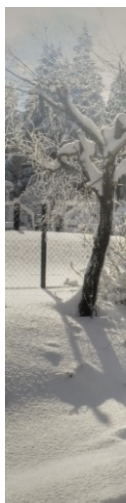


- distinguish:
 - genuine crowd-source: by lots of single users (e.g. mobile phones,...)
 - **operated networks** (partners): wind parks, photovoltaic power networks, car manufacturers, road weather stations, etc.
 - projects:
 - DWD: renewable energy
 - DWD: **car manufacturer (Audi)**
 - MCH: meteo-drones
- need to cope with data of often reduced quality (+ missing meta data):
 - minimize **conditional biases** in obs increments ($H(\mathbf{x}) - \mathbf{y}$)
by modelling of micro climate of obs device / station,
e.g. directional dependency of errors of wind obs from masts of wind parks, car microclimate, etc.
 - suitable 'situation-dependent' H and / or bias correction (of \mathbf{y})
e.g. **conditional non-linear bias correction** (Otkin + Potthast 2019, MWR), machine learning,...
 - similarly: **quality control**, specification of obs errors \mathbf{R} , time and spatial aggregation



Crowd-sourced data: motivation for project with AUDI

autonomous driving ← depends strongly on weather conditions!



Source: AUDI AG

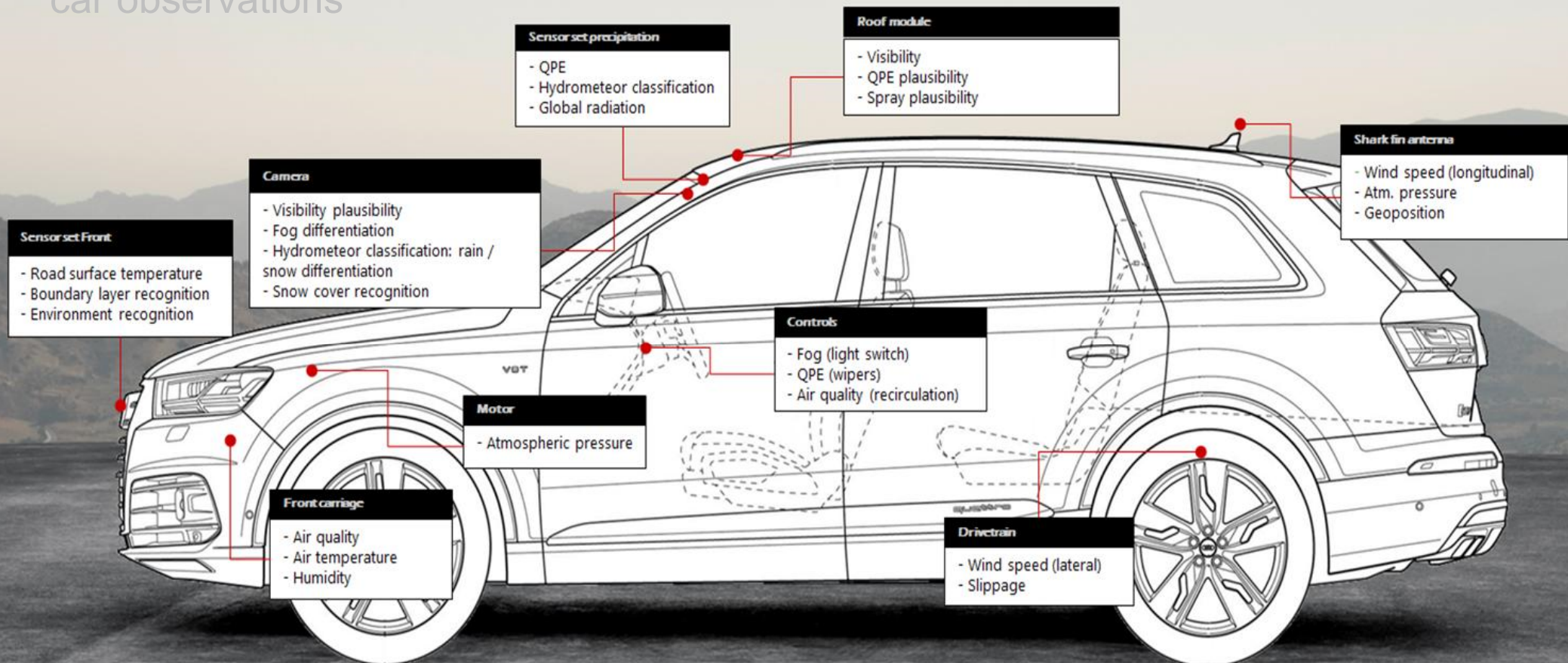
FlowKar
Flotten-Wetter-Karte

Real-time weather for a 5-60 Minutes period



Crowd-sourced data: car observations

car observations



Source: AUDI

legal issues to make data available → currently data only from 1 car
(data anonymization)

observation operators under development

road weather stations



Kombi-Sensor

erfasst Lufttemperatur, relative Feuchte, Niederschlagsintensität, Niederschlagsart, Niederschlagsmenge, Luftdruck, Windrichtung und Windgeschwindigkeit.



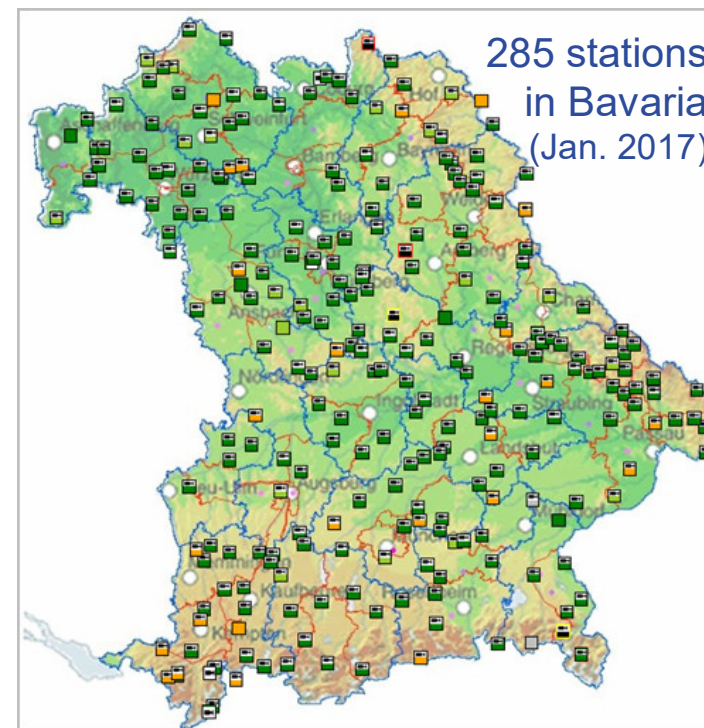
Nachtsichtfähige Kamera

ist fest installiert mit einem Infrarotscheinwerfer, der nicht den Verkehr blendet. Alle 10 Minuten wird ein Straßenbild (schwarzweiß oder farbig) erzeugt.



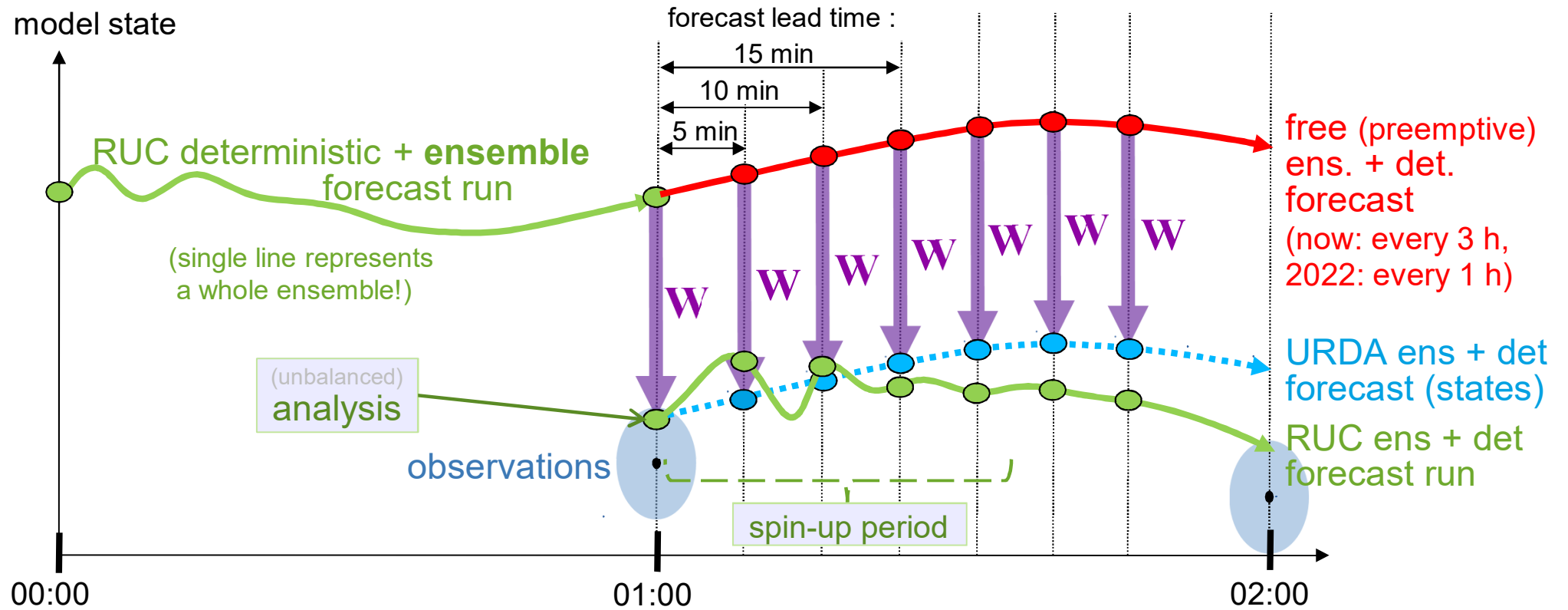
measured variables

- temperature (air + road + 30 cm depth)
- dew point
- wind (direction + velocity)
- precipitation (type + intensity)
- visibility
- road condition (dry, wet, ice, snow, ...)



Source: Bavarian Administration

Ultra Rapid Data Assimilation (URDA) for real-time weather (~ 5 – 60 min)



- RUC = Rapid Update Cycle → 1-hrly LETKF DA cycle (i.e. based on ensemble filter)
- **W** : LETKF transform matrix (defining linear combination of first guess ensemble members)

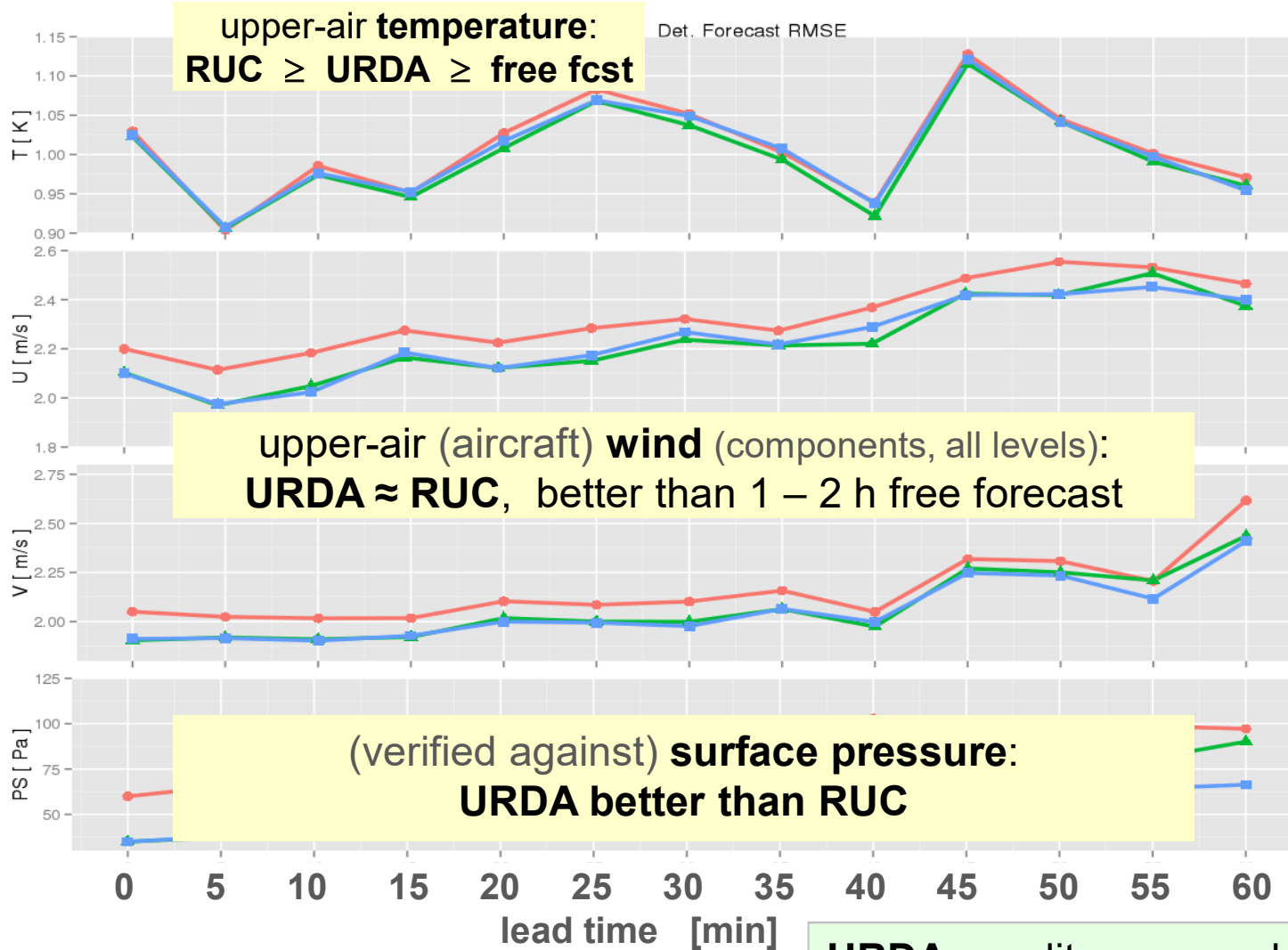
Properties:

- equivalent to standard sequential DA for **linear** model and observation operator [Potthast & Welzbacher 2018]:
 - URDA applicable only for (very) **short lead times** ($\leq \sim 1 - 3$ h) for which
 - non-linearity is sufficiently 'small'
 - local transform matrix (of LETKF) still applicable to the local forecast (i.e. the influence of obs remain within localisation area)
- applicable to **several** assimilation steps from successive time intervals
- no model re-initialization / **no additional forecast runs** necessary
 - no need to update the whole model state;
a **reduced set** of selected variables and grid points can be updated
 - **very low computational costs**
enables forecasts updated very **frequently + timely**

Ultra Rapid Data Assimilation (URDA) experiment (single case!)

RMSE vs. lead time (period 60 minutes)

assimilated: Synop pressure
radiosonde
aircraft
wind profiler data



URDA: quality comparable to RUC, no spin-up effects
→ great potential for very short lead times



- **3-D LETKF** almost as good as **4-D LETKF**
- **ICON-LAM** much better than COSMO
- **3-D radar reflectivity** in LETKF beats Latent Heat Nudging (ARPAE)
- small extra benefit from (low-elevation) **GPS STD** on top of ZTD
- work on **crowd-sourced** data e.g. related to cars – new obs operators, bias correction, ...
- (ensemble-based) **Ultra Rapid DA** (URDA) promising for real-time weather

Thank you for your attention

